
Some Recent Advances in Algebraic Multigrid

Parallel AMG and AMGe (element)

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Parallel AMG requires parallel algorithms for these steps:

- *The Setup Phase*
 - Coarse Grid Selection
 - Construction of Prolongation operator, P
 - Construction of coarse-grid operators by Galerkin method, $P^T A P$
- *The Solve Phase*
 - Residual Calculation
 - Relaxation
 - Prolongation
 - Restriction

Parallelizing the *Solve* Phase

- *The Solve Phase*

- **Residual Calculation**

- entails Matvec: $y \leftarrow \alpha A x + \beta y$

- **Relaxation**

- We use Jacobi rather than Gauss-Seidel
 - entails scaled Matvec-like operation

$$x \leftarrow x + D^{-1}(b - Ax)$$

Parallelizing the *Solve* Phase, II

- *The Solve Phase*

- **Prolongation**

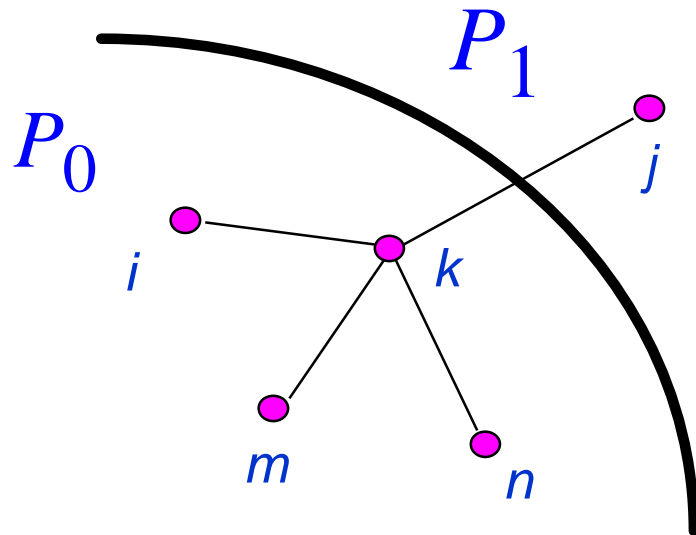
- Requires a simple Matvec, but on a rectangular matrix. May not be available in some common toolkits, but readily built and easily parallelizable

- **Restriction**

- Requires a MatvecT, the product of the transpose of a rectangular matrix. Not generally available in toolkits, but is easily constructed.

The Parallel *Setup* phase

- **Construction of Prolongation operator, P** , requires “processor boundary” equations (ghost point information), and can be accomplished using toolkit functions.



P_0 requires:

Row i

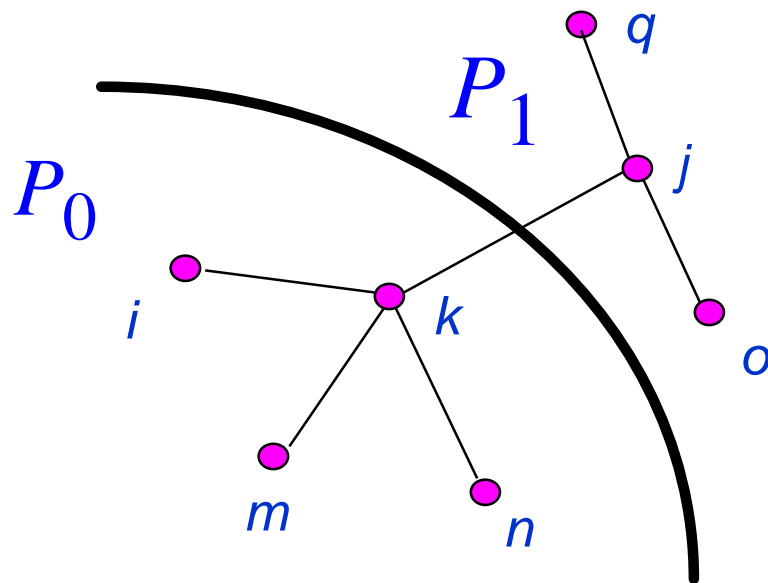
Row m

Row n

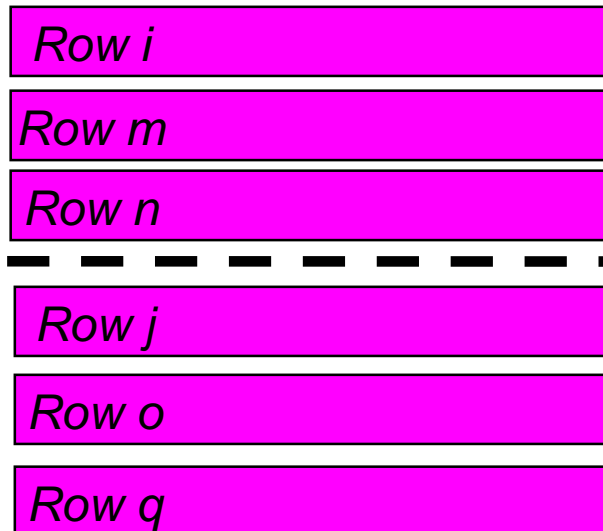
Row j

The Parallel *Setup* phase

- **Construction of coarse-grid operators** by Galerkin method, $P^T A P$, requires **two** layers of processor boundary data.



P_0 requires:



The Parallel *Setup* phase, II

- **Selection of the coarse-grid points**
 - The *main challenge of parallelizing AMG*. The standard algorithm is inherently serial, requiring pathlength-two updates after each C-point is selected before work can continue.
 - We have **developed a parallel coarsening algorithm** that uses a Luby-Jones-Plassman like MIS algorithm to select coarse-grid points based on an “influence measure” that favors points that influence many other points.

Useful definitions

- The variable u_i **depends** on the variable u_j if the j th coefficient in the i th equation is large compared to the other off-diagonal coefficients in the i th equation. That is, if
$$-a_{ij} > \theta \max_{j \neq i} (-a_{ij}) \quad (\text{assumes M-matrix})$$

- If j **depends** on k , then k **influences** j , which is denoted graphically by:



- The set of coarse-grid variables is denoted C .
- The set of coarse-grid variables used to interpolate the value of the fine-grid variable u_i is denoted C_i .

S , the matrix of influence and dependence

- Define S to be the adjacency matrix of the graph of influence associated with the operator, A .

$$S_{ij} = \begin{cases} 1 & i \Rightarrow j \\ 0 & \text{else} \end{cases} \quad \begin{matrix} (i \text{ depends on } j \\ j \text{ influences } i) \end{matrix}$$

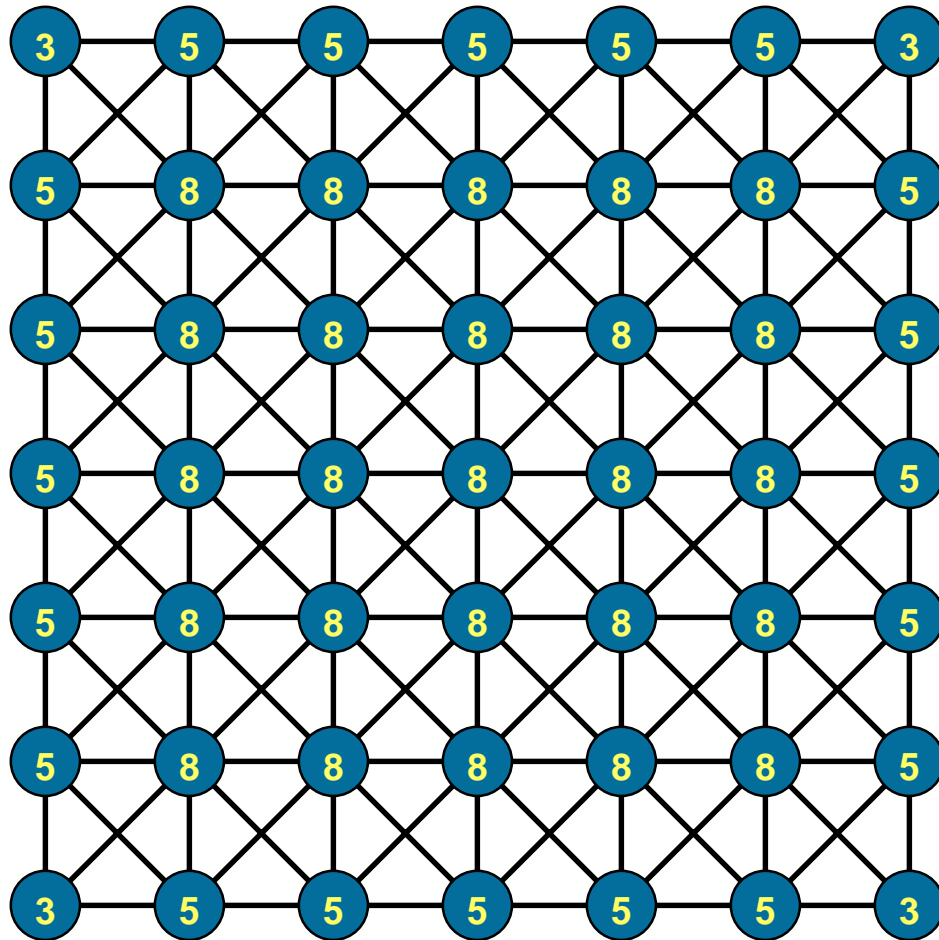
- The nonzero columns in the i th row of S , denoted $S_{i:}$, form the set of **dependencies** of the point i .
- The nonzero rows in the i th column of S , denoted $S_{:i}$, form the set of **influences** of the point i .

Standard AMG:

Choosing the Coarse Grid

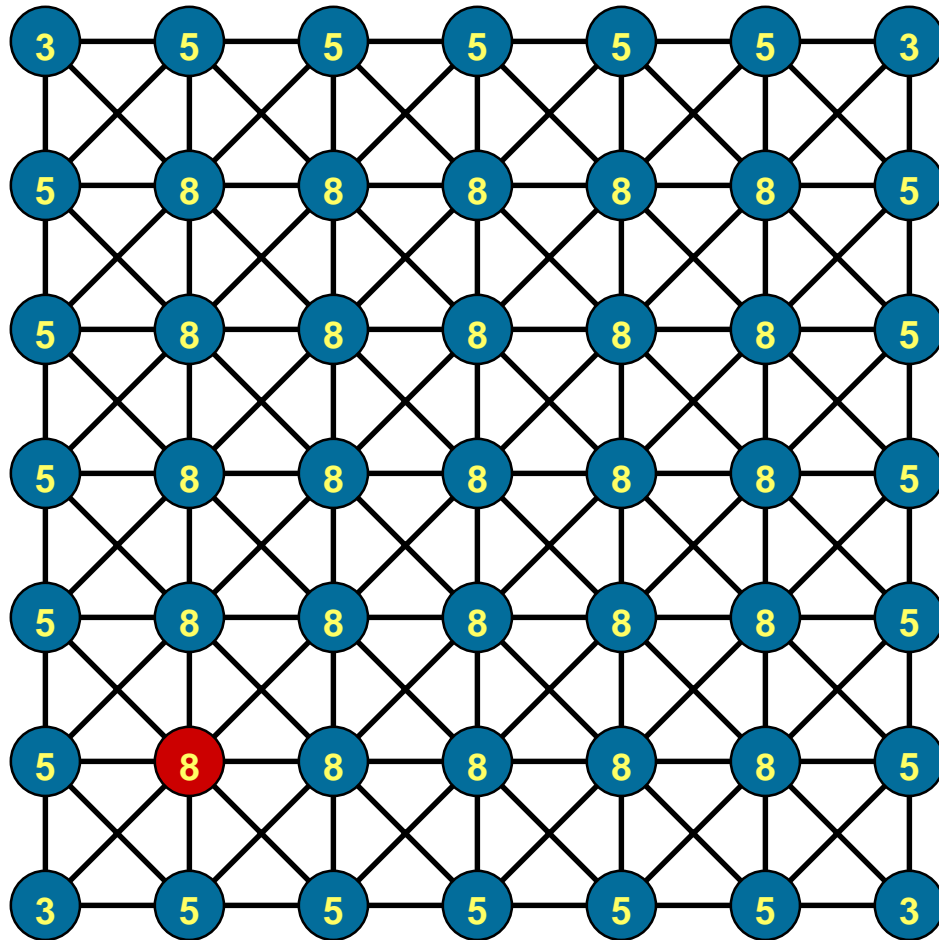
- **Two Criteria**
 - **(C1)** For each fine-grid point i , each $j \in S_i$: should either be in C or should be dependent on at least one point in C_i .
 - **(C2)** C should be a **maximal subset** with the property that no C -point influences another C -point.
- Satisfying both (C1) and (C2) is sometimes impossible. We use (C2) as a guide while enforcing (C1).

Standard AMG coarsening: algorithm is **inherently** sequential



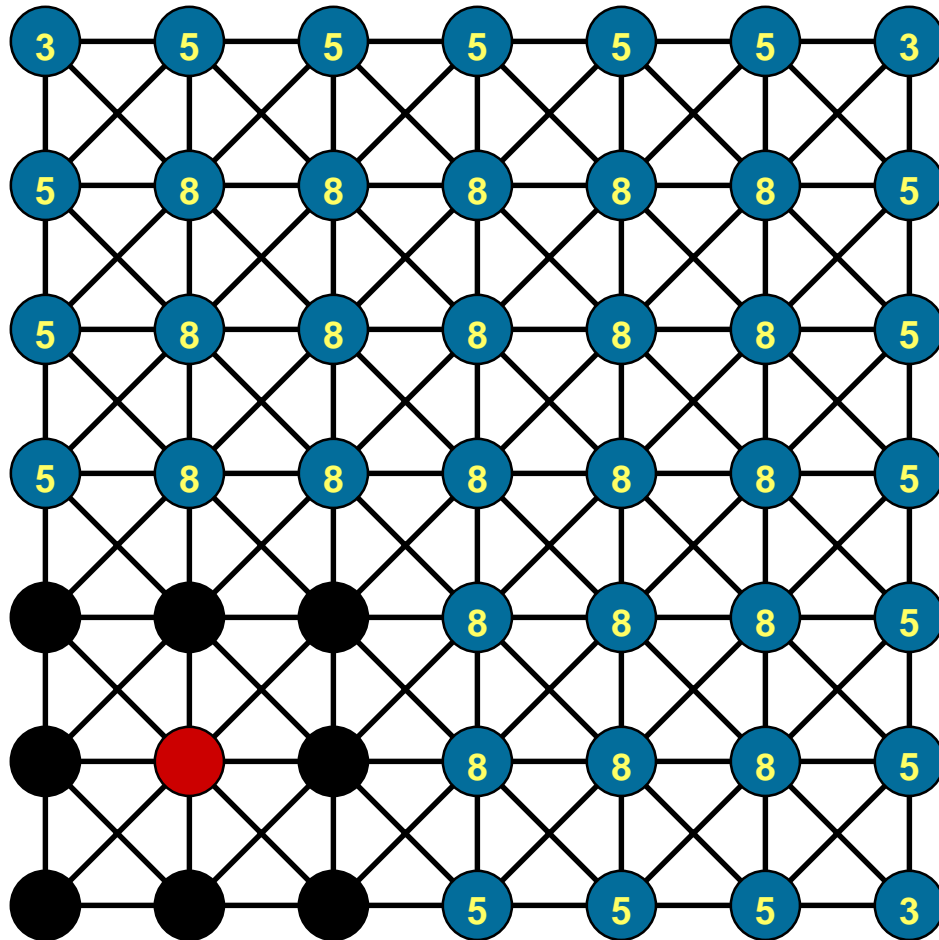
- ➡ select C-pt with maximal measure
- ➡ select neighbors as F-pts
- ➡ update measures of F-pt neighbors

Standard AMG coarsening: algorithm is **inherently** sequential



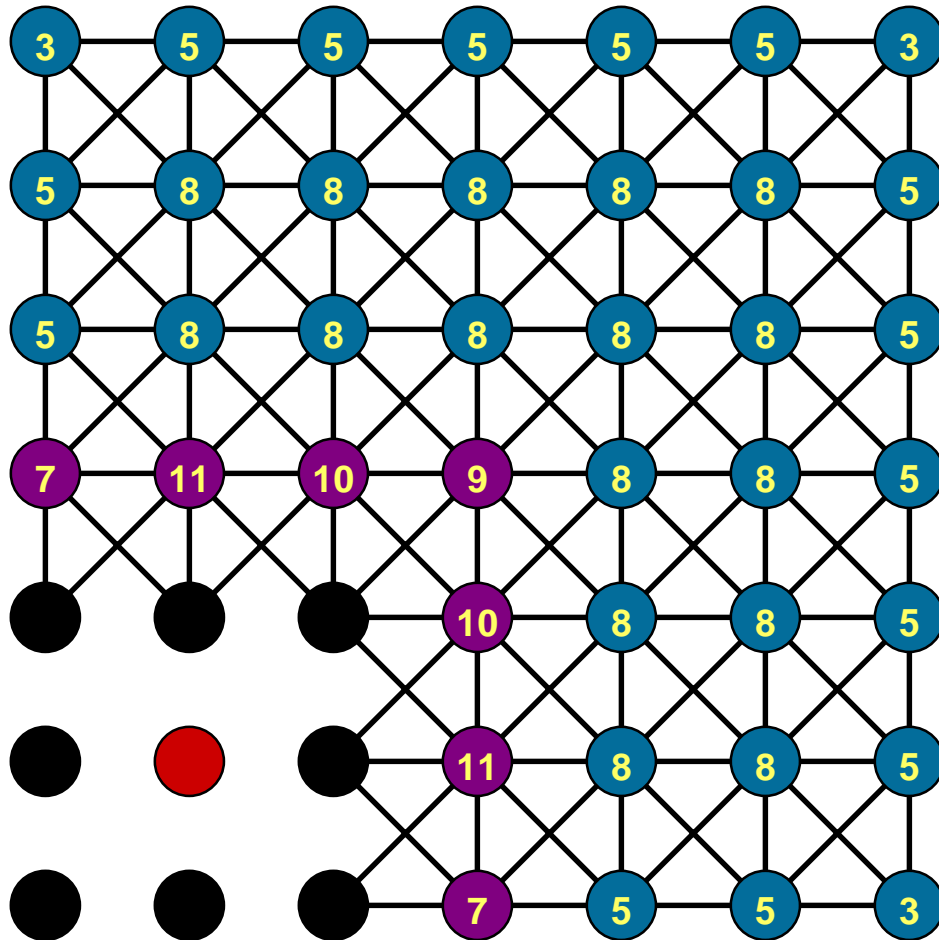
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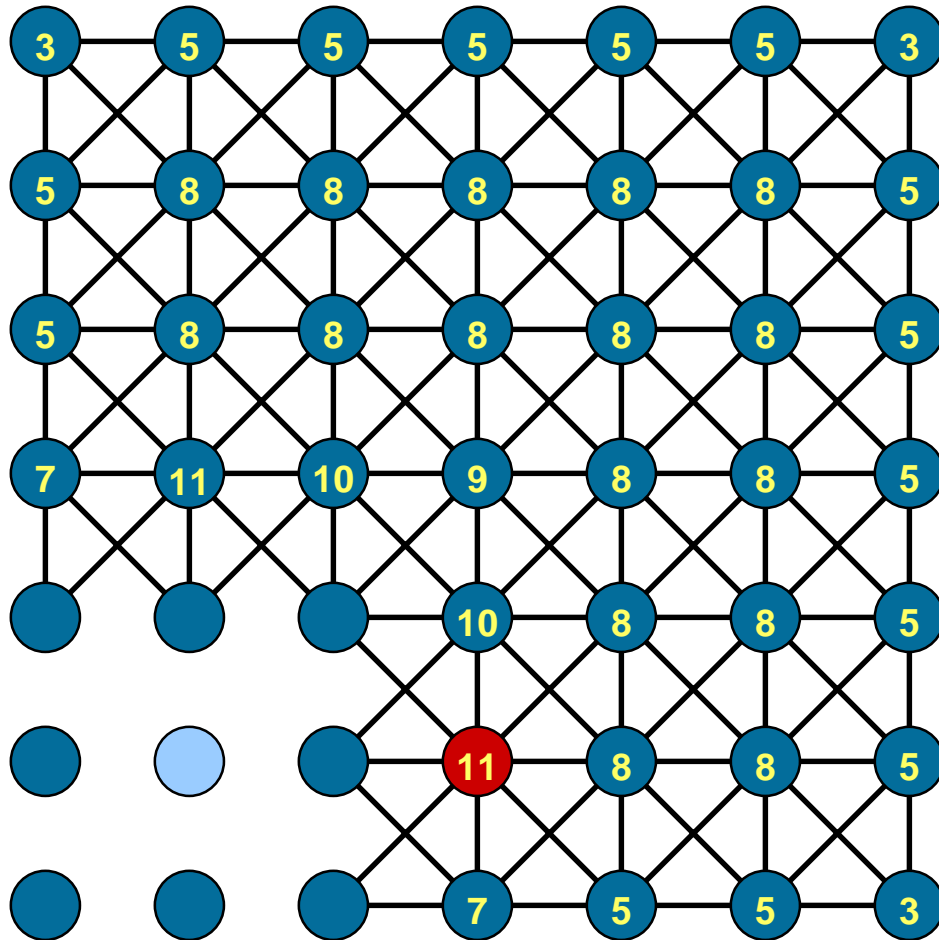
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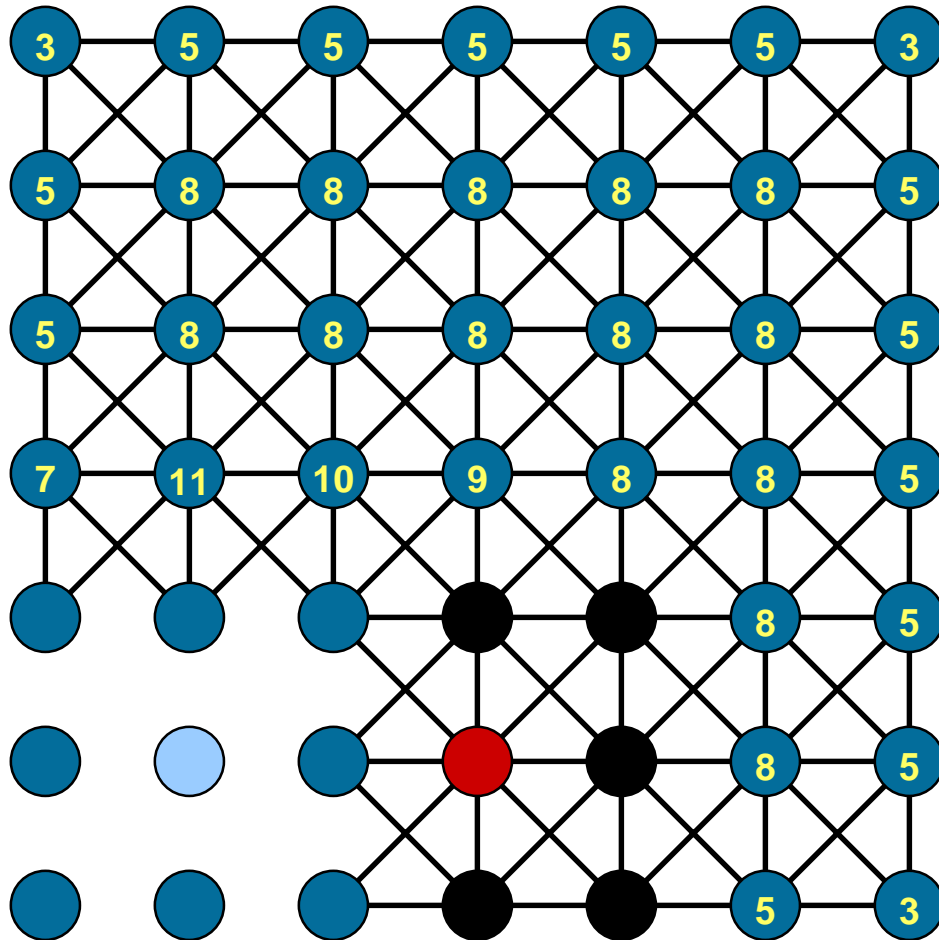
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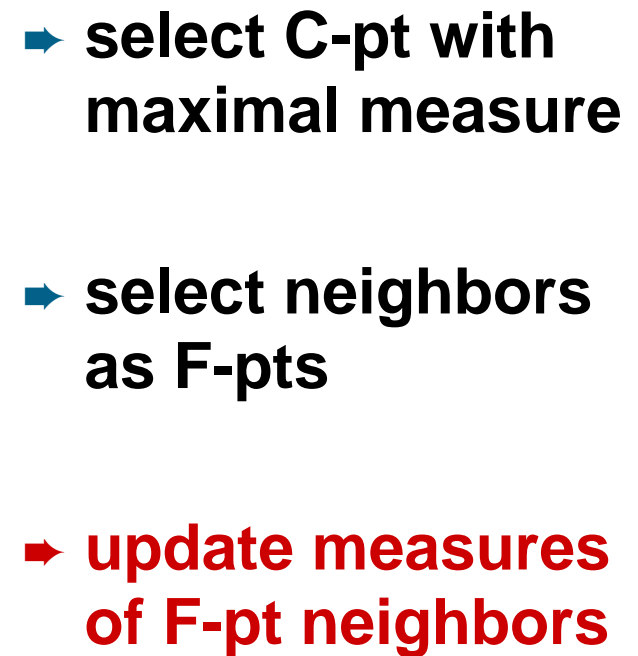


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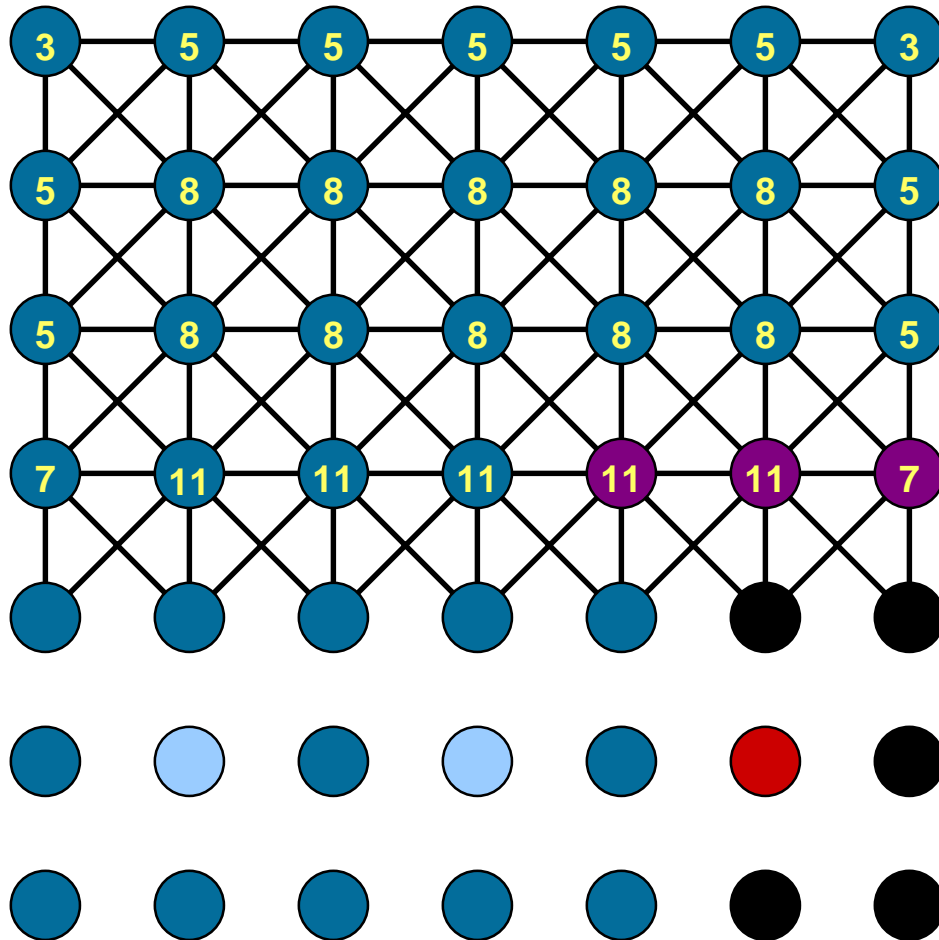
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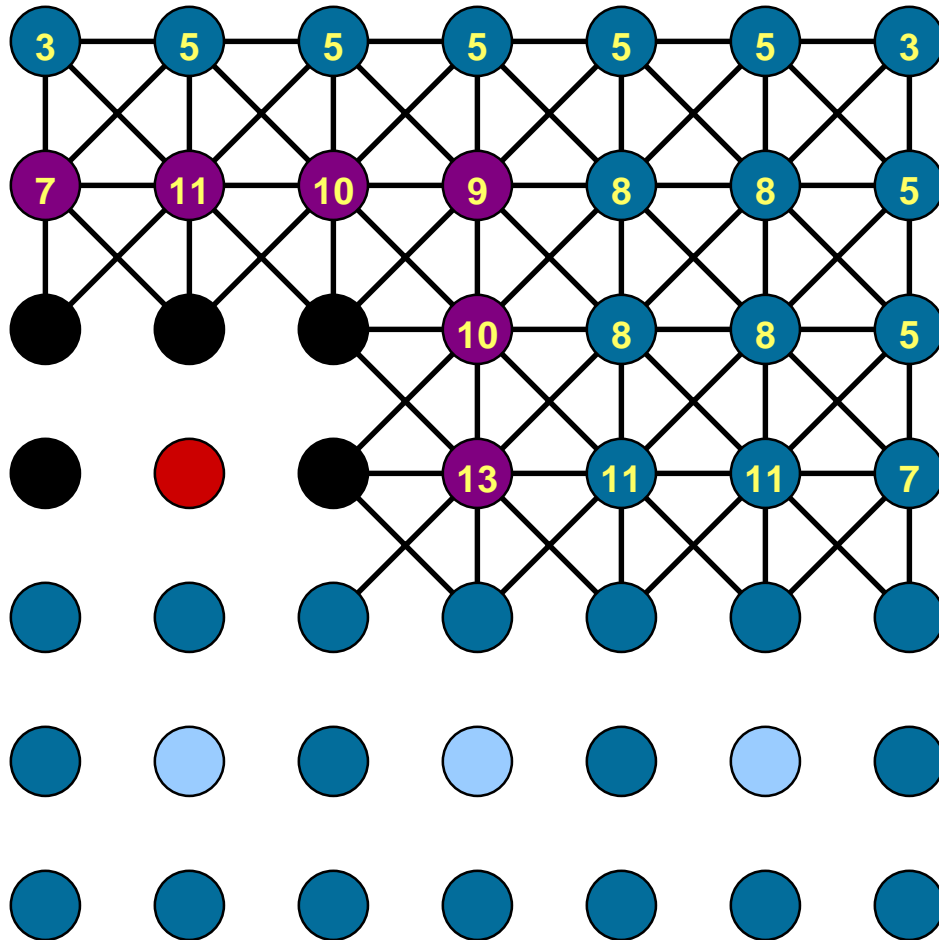


Standard AMG coarsening: algorithm is **inherently** sequential



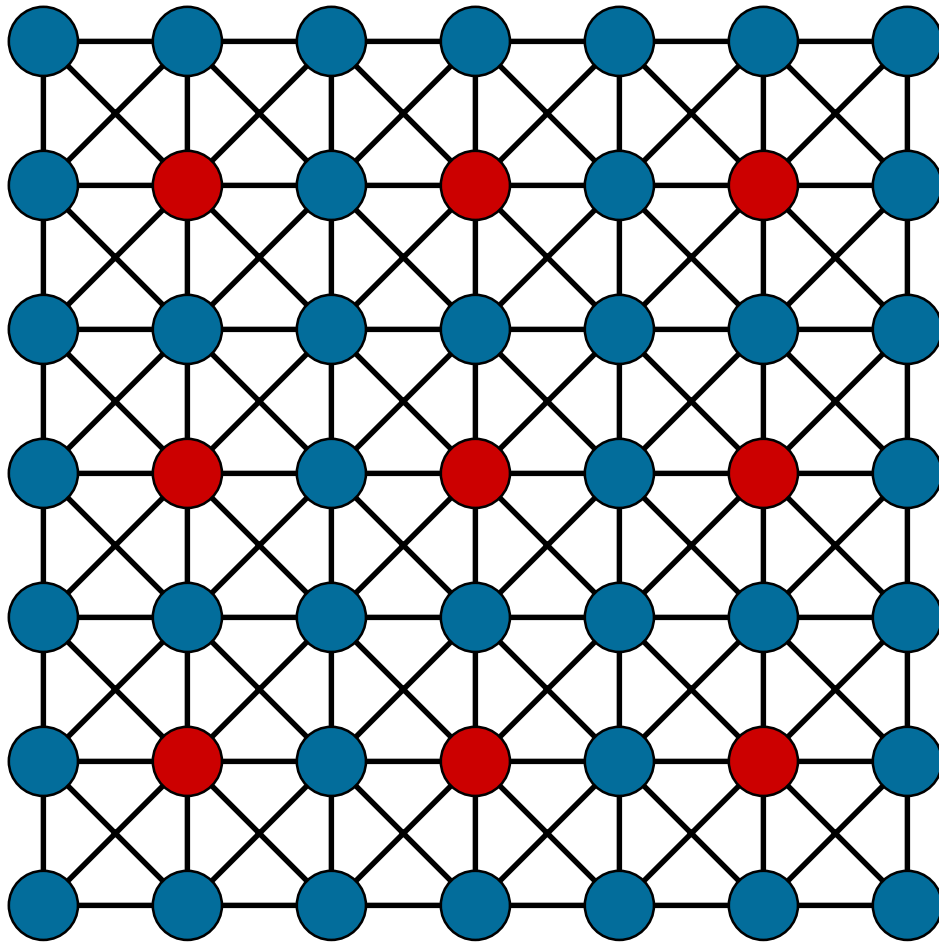
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ParAMG Coarsening

- Create a “measure” of each point, consisting of the number of influences of the point, plus a random number in $[0,1]$.
- Select a set of points whose measure exceeds that of all points they **influence or depend on**. Set is **independent** by construction, may be maximal, (needn't be). **This can readily be done in parallel, since once the random values are distributed, the only action is read-only!**
- Perform **ParAMG** heuristics (described below) on the set of points selected above. Can be done in distributed fashion, requiring a synchronization at the end of the step when the set is exhausted.

ParAMG Coarsening Heuristic 1: the effect of coarse points

- The values at a C-point is not interpolated, hence C-points won't need to interpolate from neighbors they depend on. Those neighbors have lessened “value” as potential C-points themselves.
 - For each neighbor, j , that influences c :
 - subtract 1 from **measure**[j]; and
 - remove the edge S_{cj} from the graph

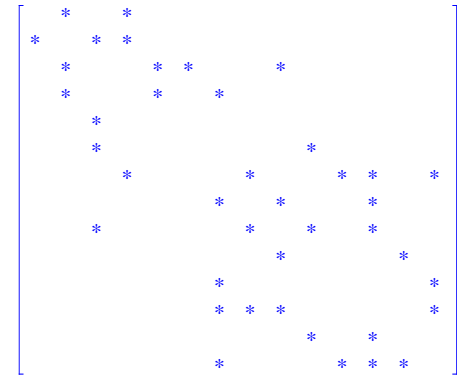
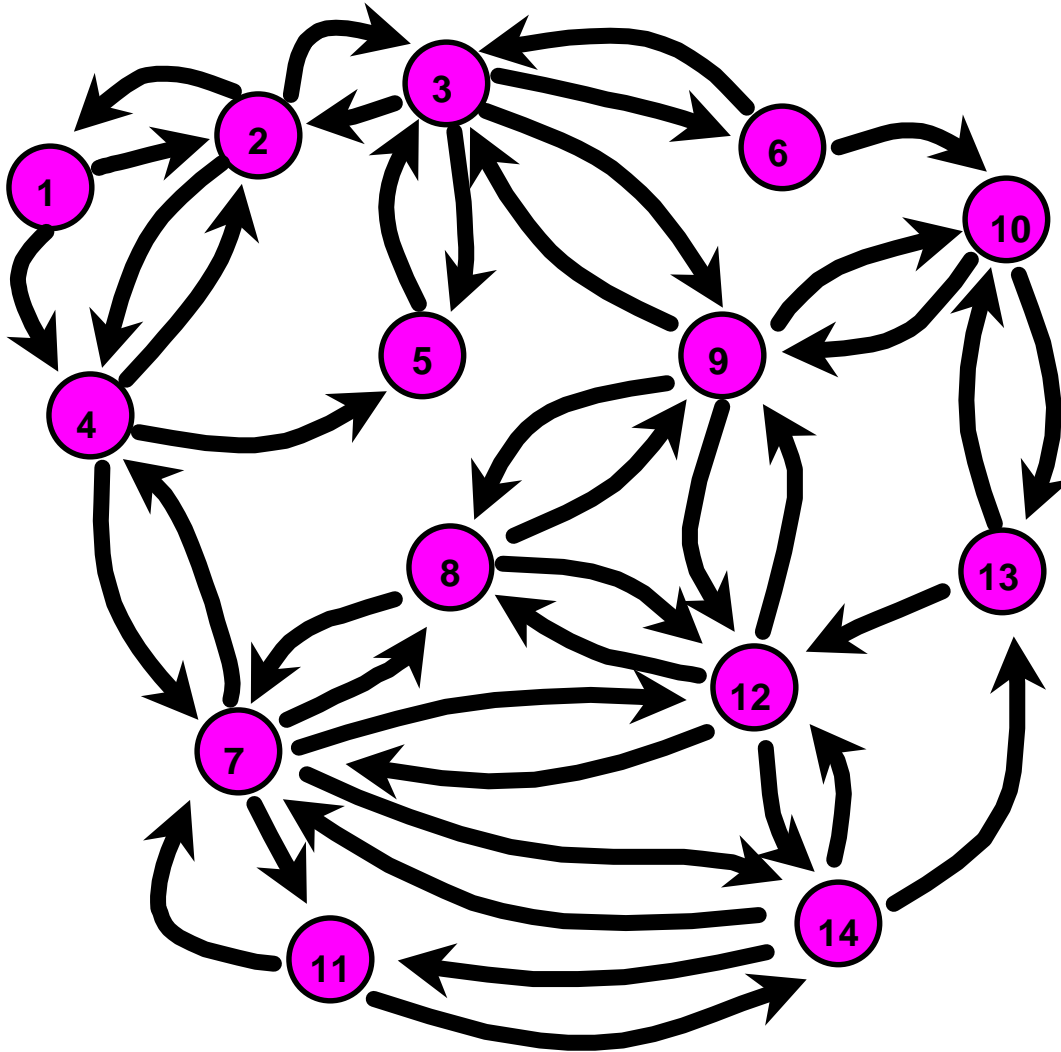
ParAMG Heuristic 2: neighbors dependent on a common C-point

- If k and j both depend on a given **C**-point, and j influences k , then the value of j as a coarse point is lessened, since k can be interpolated from C .
 - For each j that C influences: (i.e., $S_{jC} \neq 0$)
 - delete S_{jC}
 - for each k that j influences:
 - if k depends on C :
 - subtract 1 from **measure[j]**;
 - remove edge S_{kj}

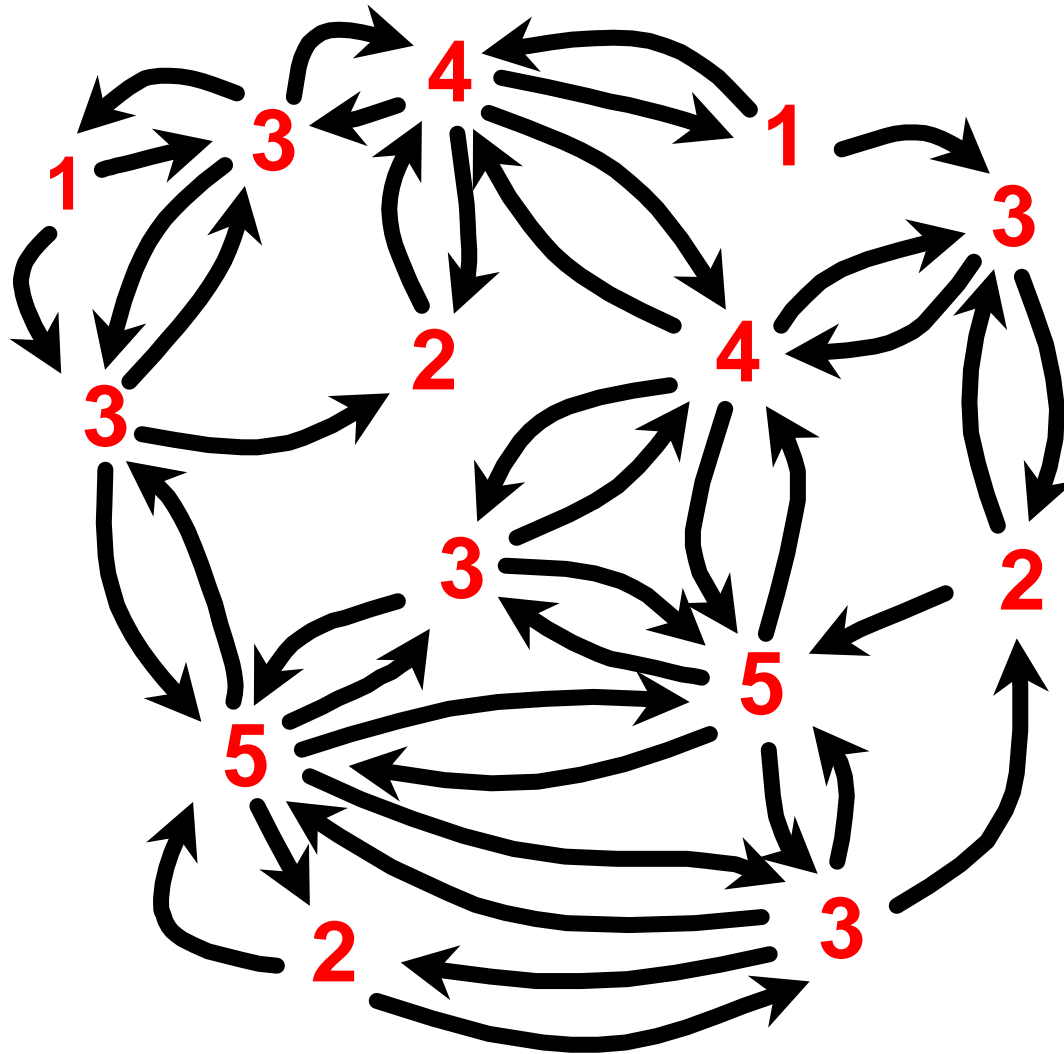
ParAMG Coarsening

- Repeat the process on those vertices and edges remaining in the graph. A vertex is removed when all its edges are removed.
- The process continues until all points have either been selected as a C-point by the independent set picker or have been removed from the graph by virtue of $\text{measure}[j] \rightarrow 0$.
- The union independent sets is the coarse grid. All other points form the fine grid.

Selecting the coarse-grid points

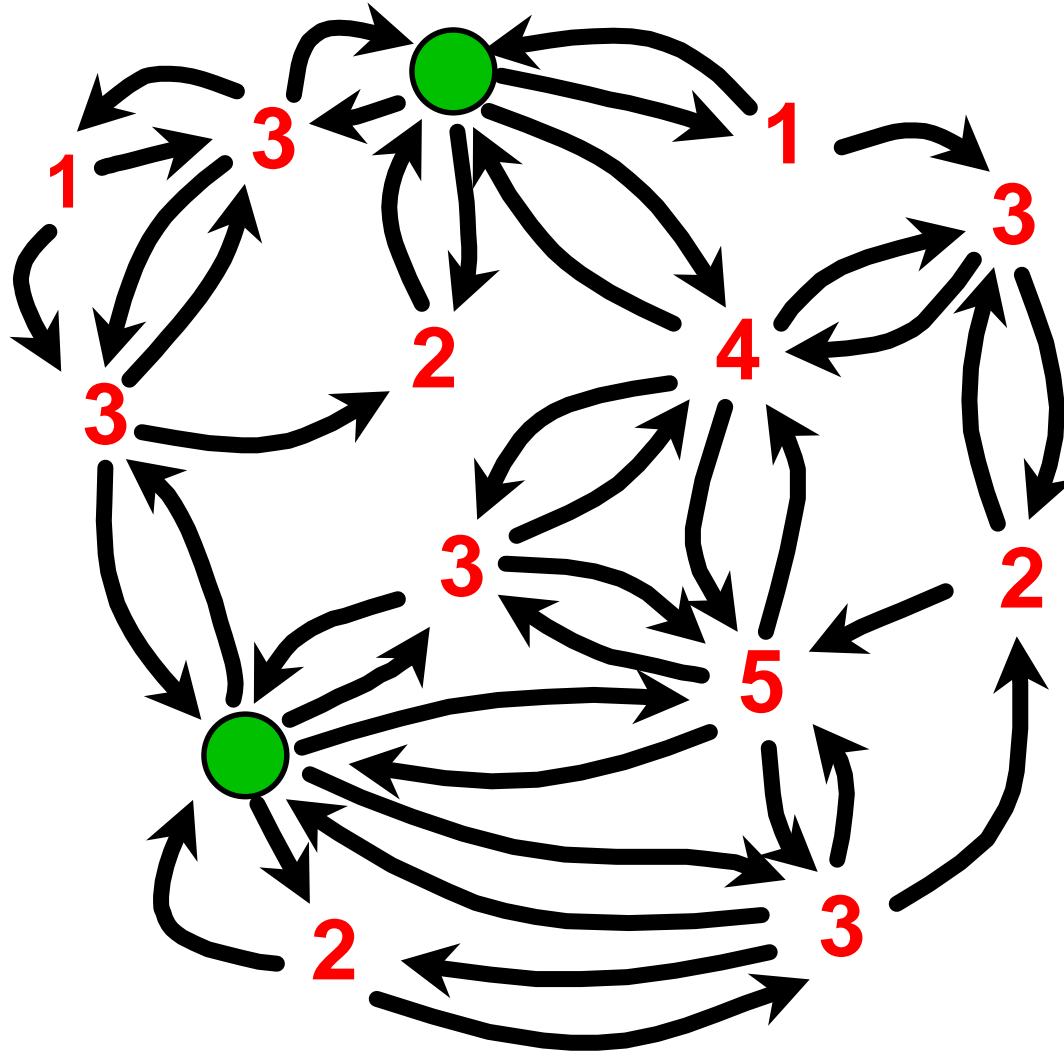


Selecting the coarse-grid points



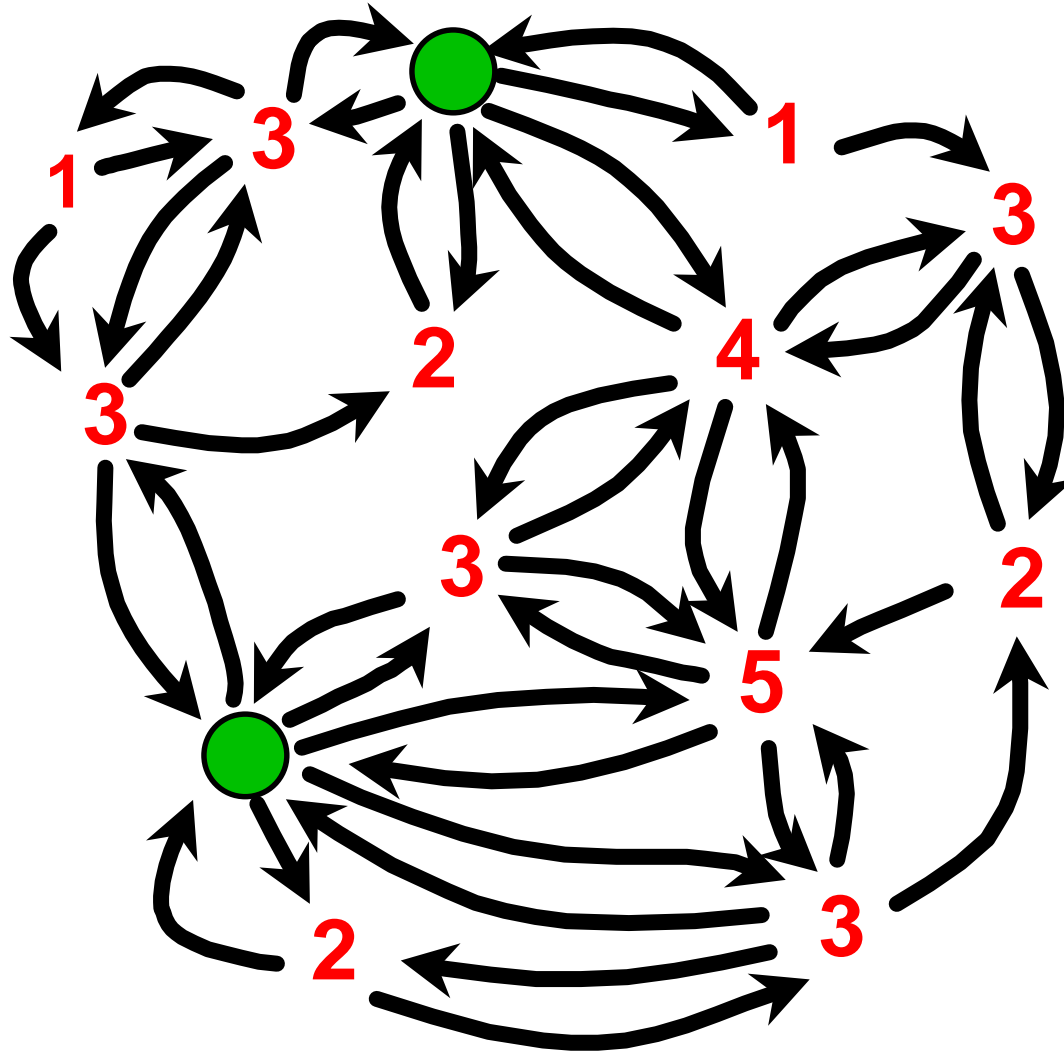
Determine the “measure” for each point: the number of other points influenced (arrows pointing into the point) plus a random number between 0 and 1 (not shown).

Selecting the coarse-grid points



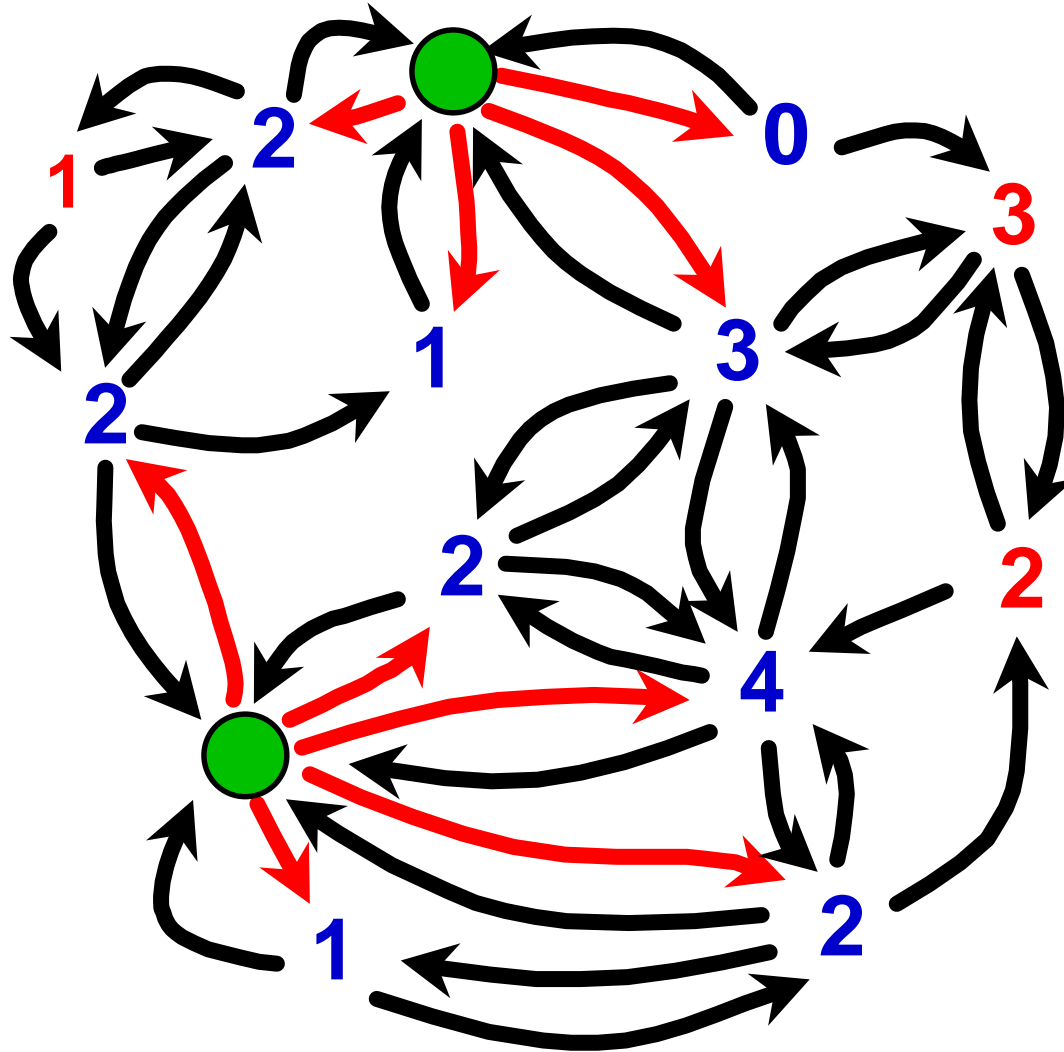
Choose as C-points an independent set of points whose measures are greater than those of all their neighbors.

Selecting the coarse-grid points



C-points will not be interpolated.

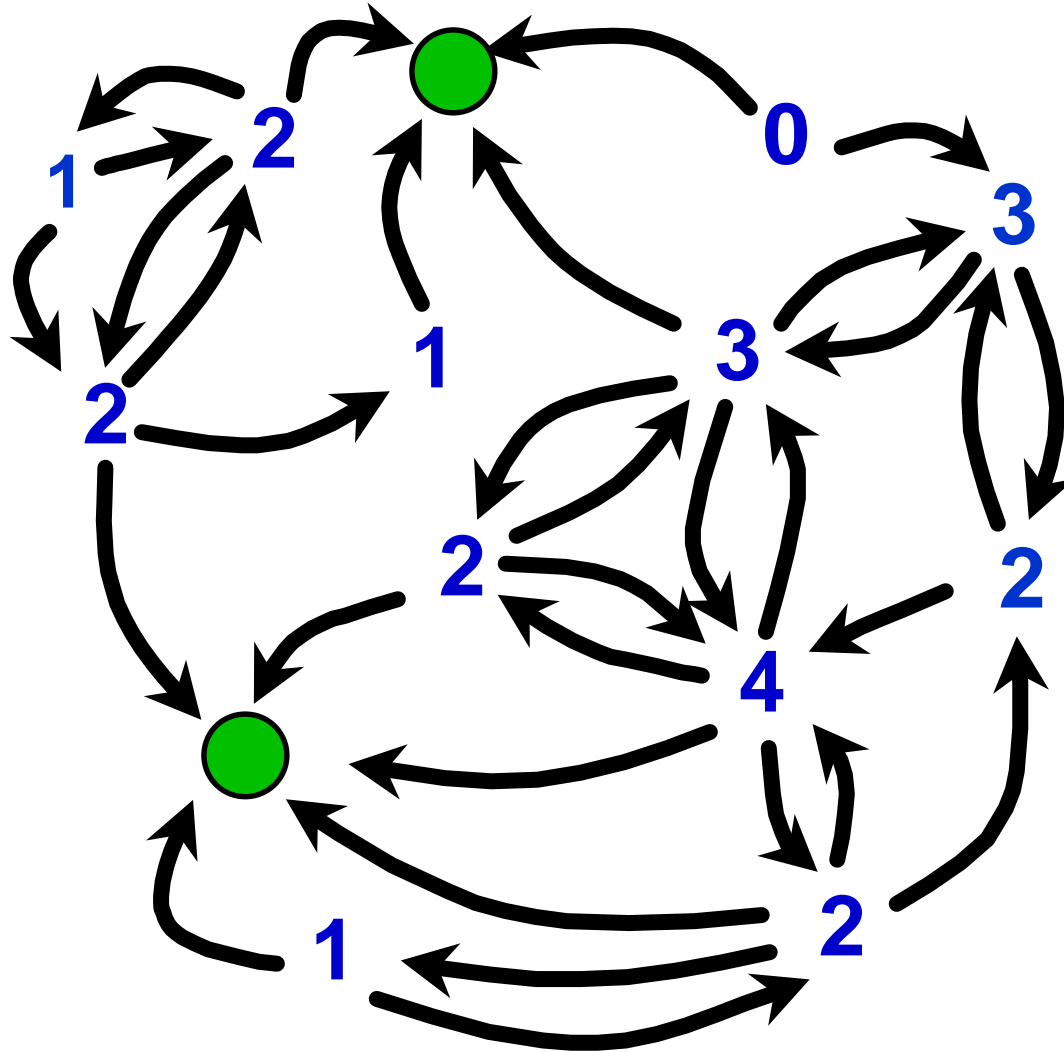
Selecting the coarse-grid points



C-points will not be interpolated.

Lower the measures of the points that influence these C-points.

Selecting the coarse-grid points

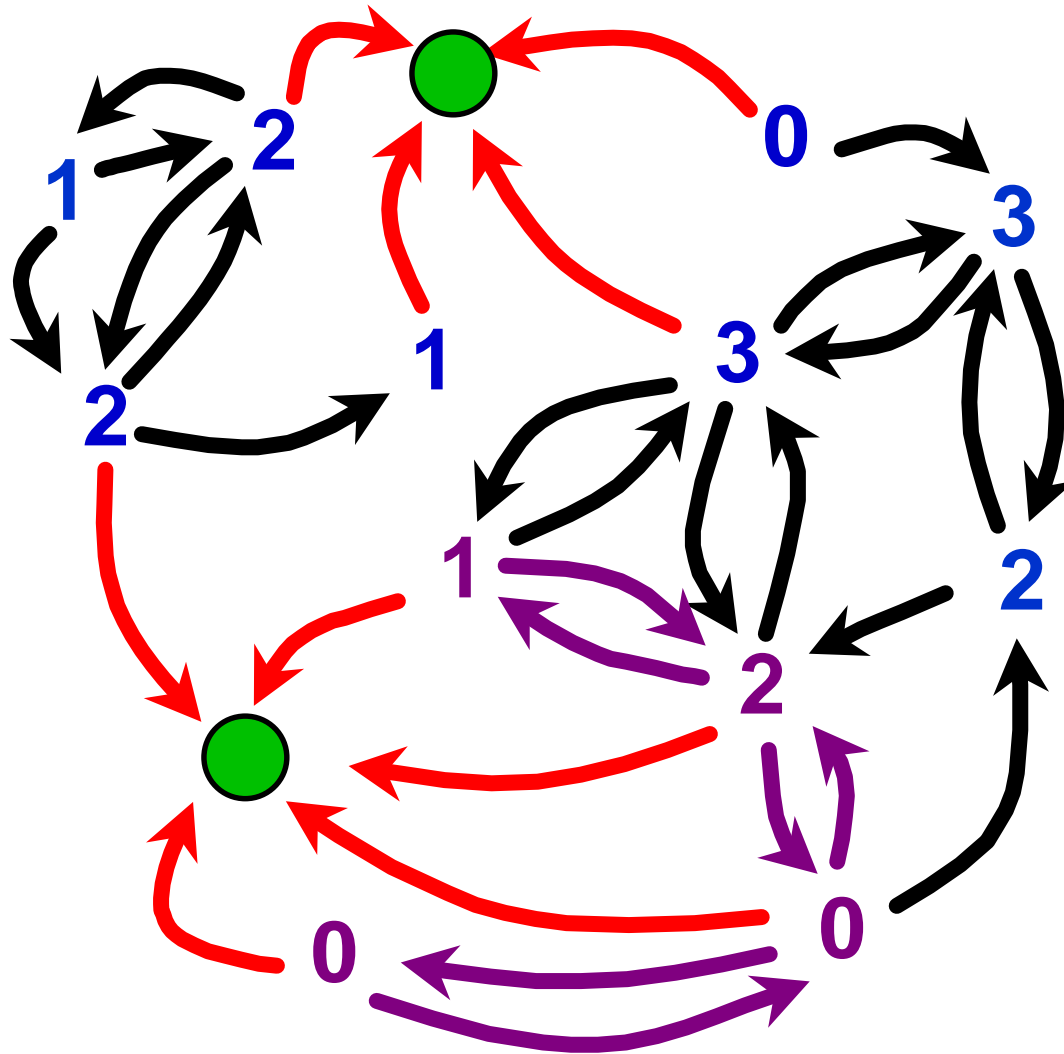


C-points will not be interpolated.

Lower the measures of the points that influence these C-points.

Remove the edges showing this influence from the graph.

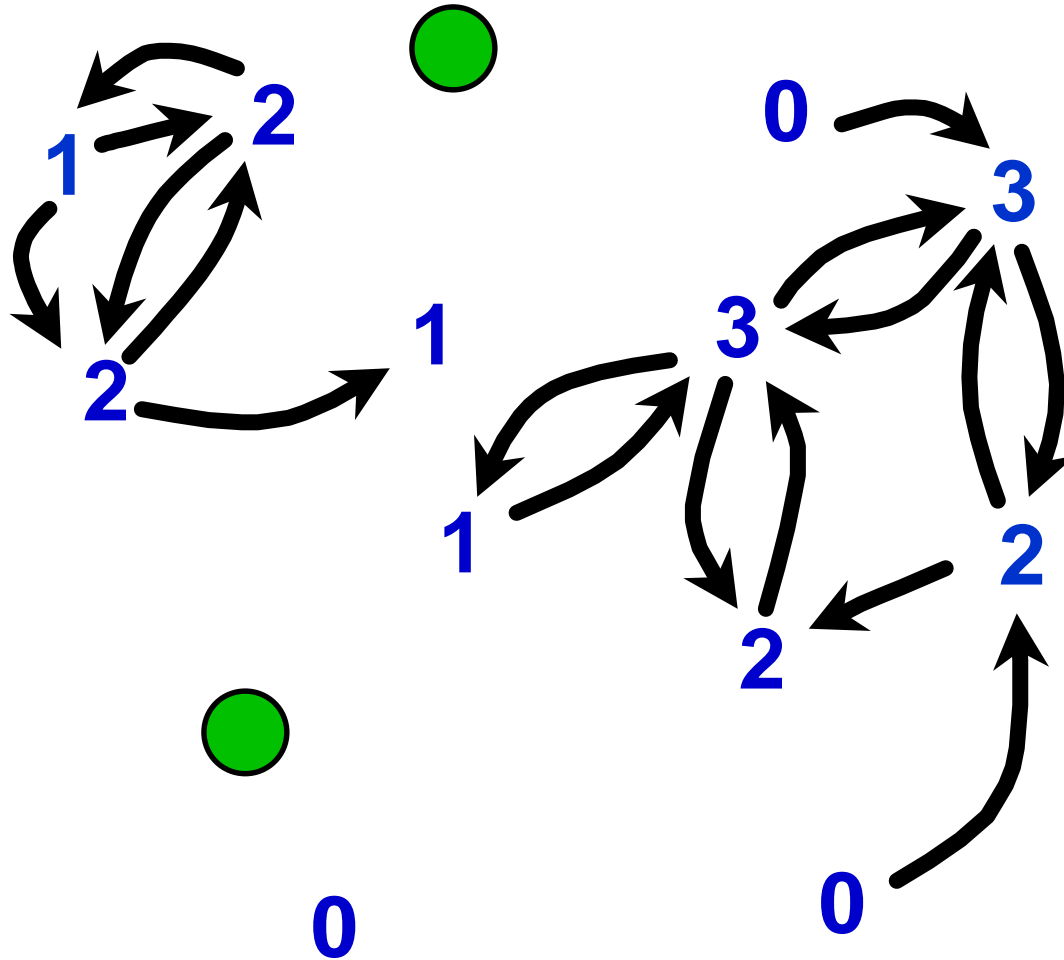
Selecting the coarse-grid points



F-points influenced by a common C-point don't interpolate each other:

Lower the measure of each point P, influenced by C, for every other point P influences that also depends on C

Selecting the coarse-grid points

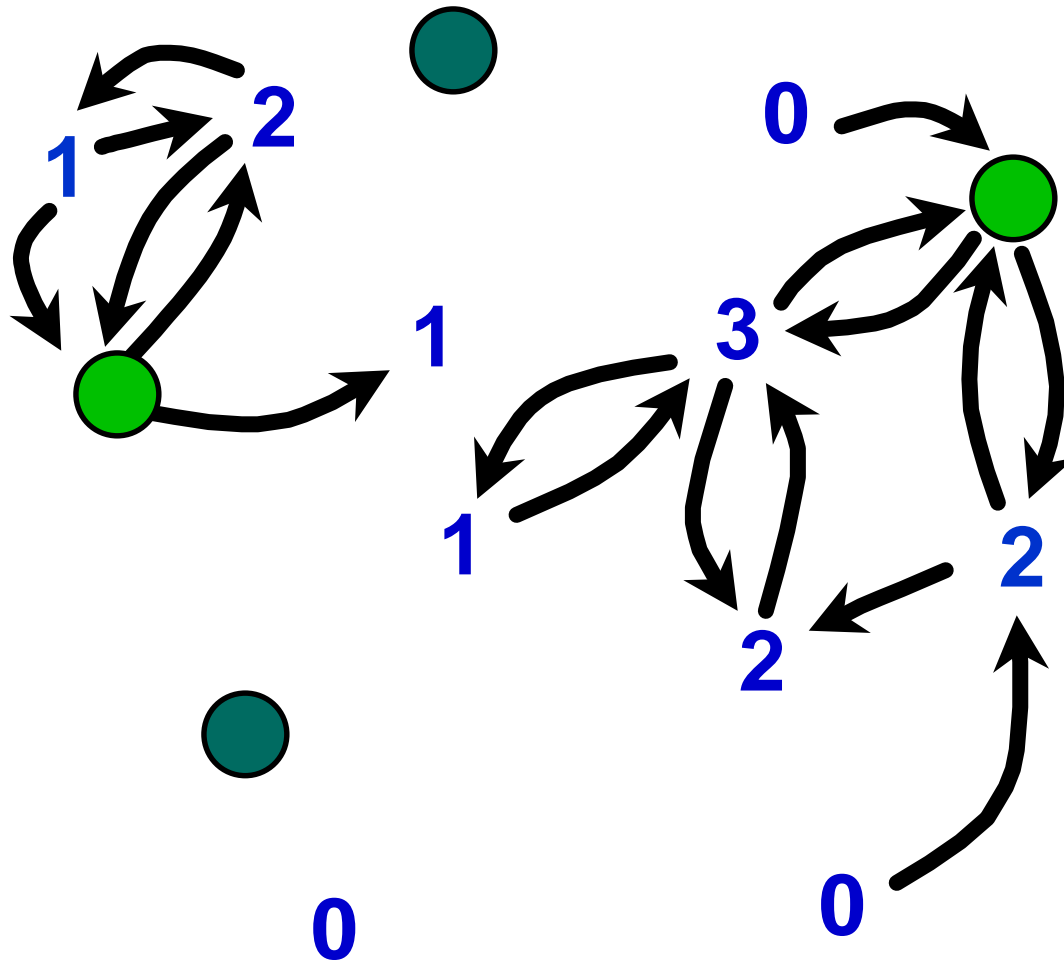


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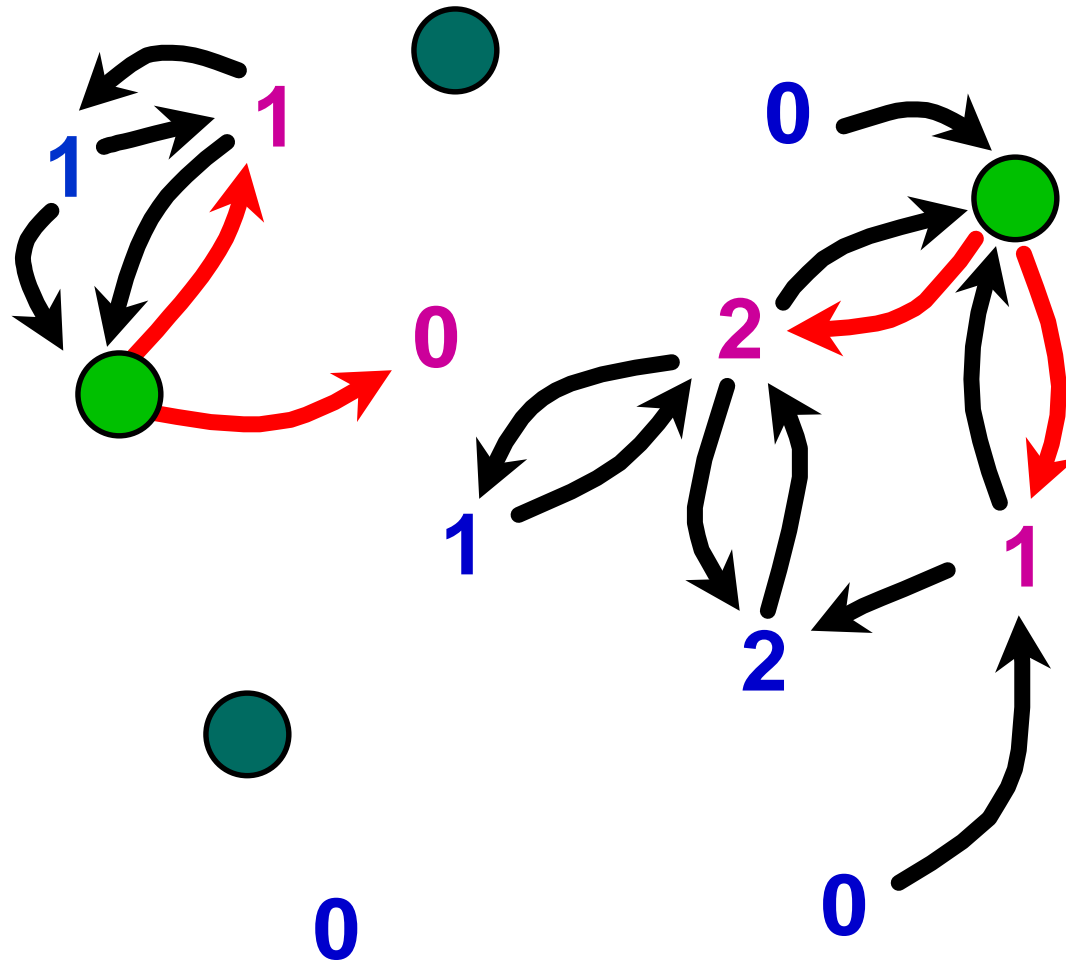
Remove the edges showing this influence from the graph.

Selecting the coarse-grid points



From the graph that remains
Choose as C-points an independent set of points whose measures are greater than those of all their neighbors.

Selecting the coarse-grid points

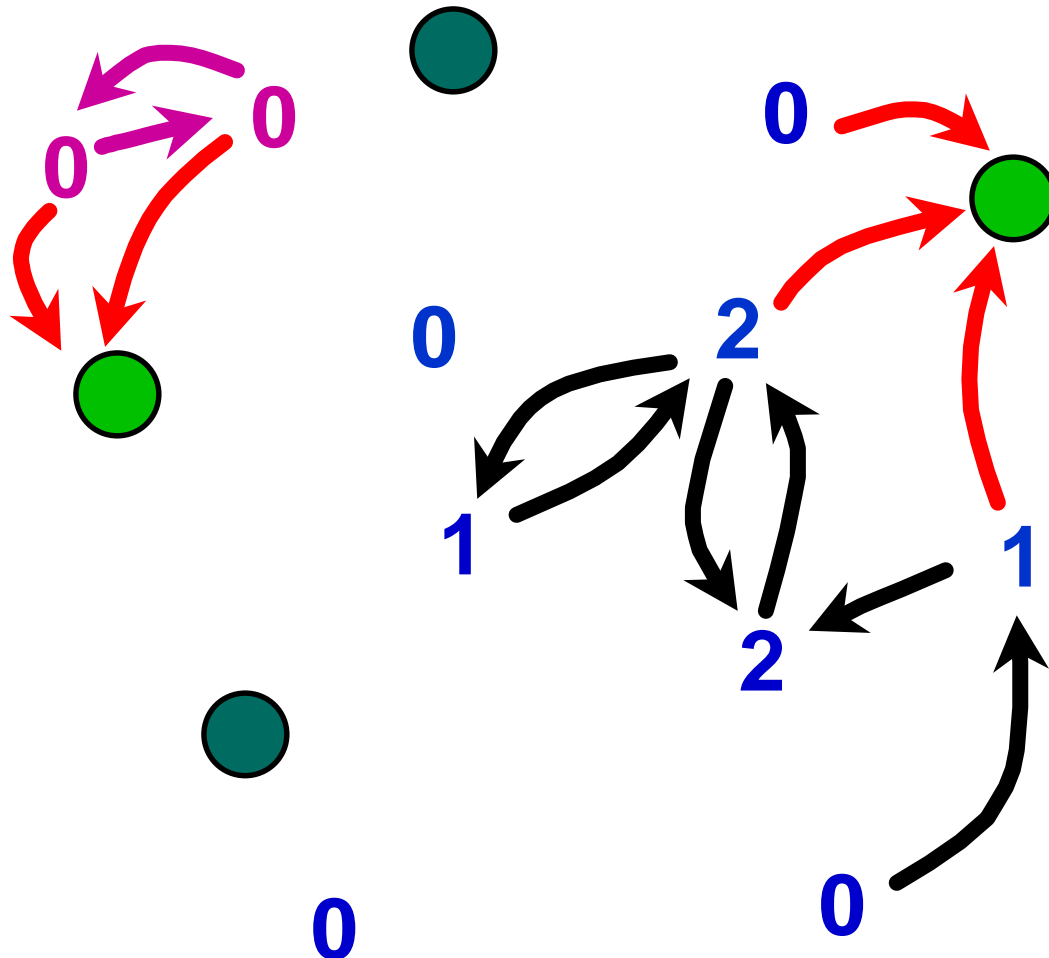


C-points will not be interpolated.

Lower the measures of the points that influence these C-points.

(Next, remove the edges showing this influence from the graph.)

Selecting the coarse-grid points

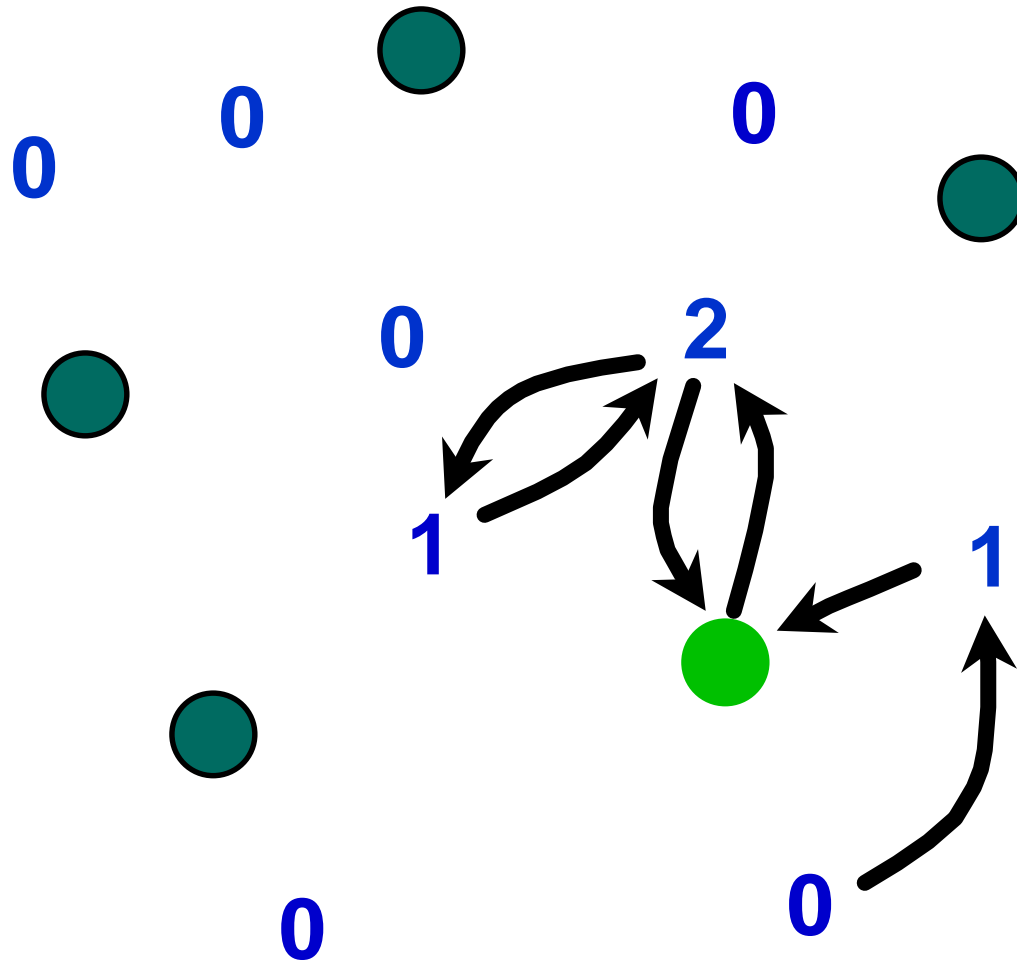


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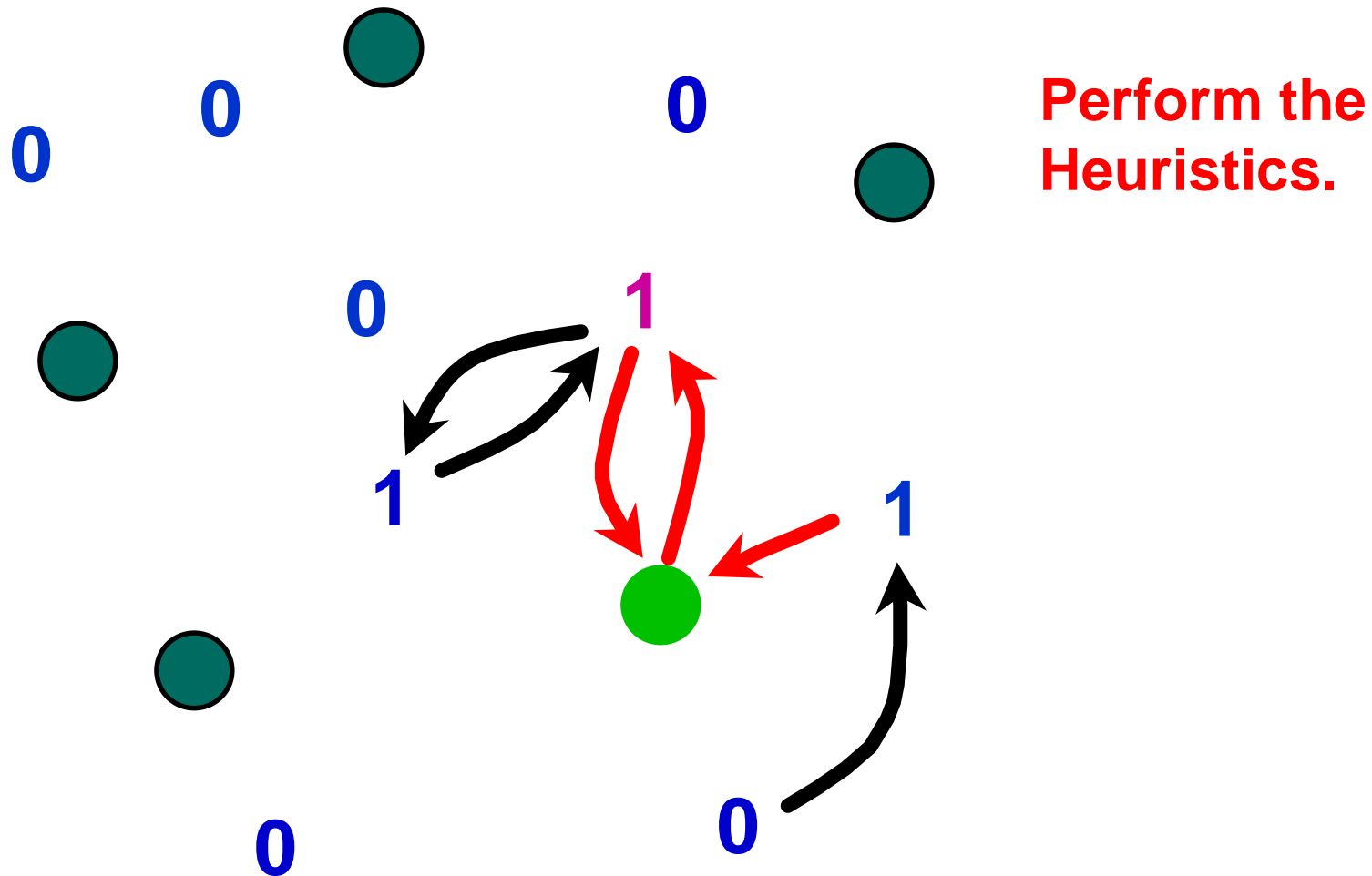
(Next, remove the edges showing this influence from the graph.)

Selecting the coarse-grid points

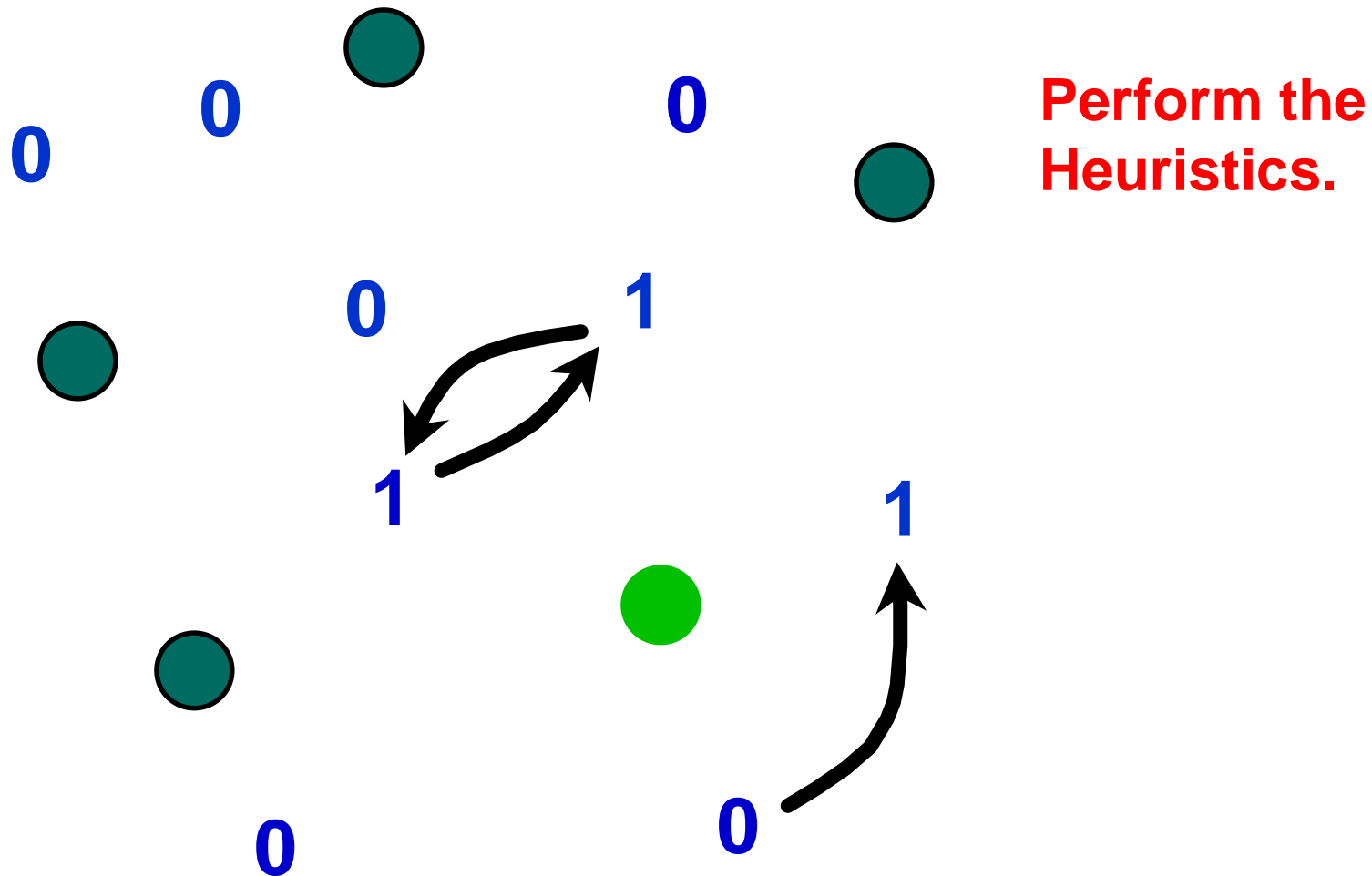


From the graph that remains
Choose as C-points an
independent set
of points whose
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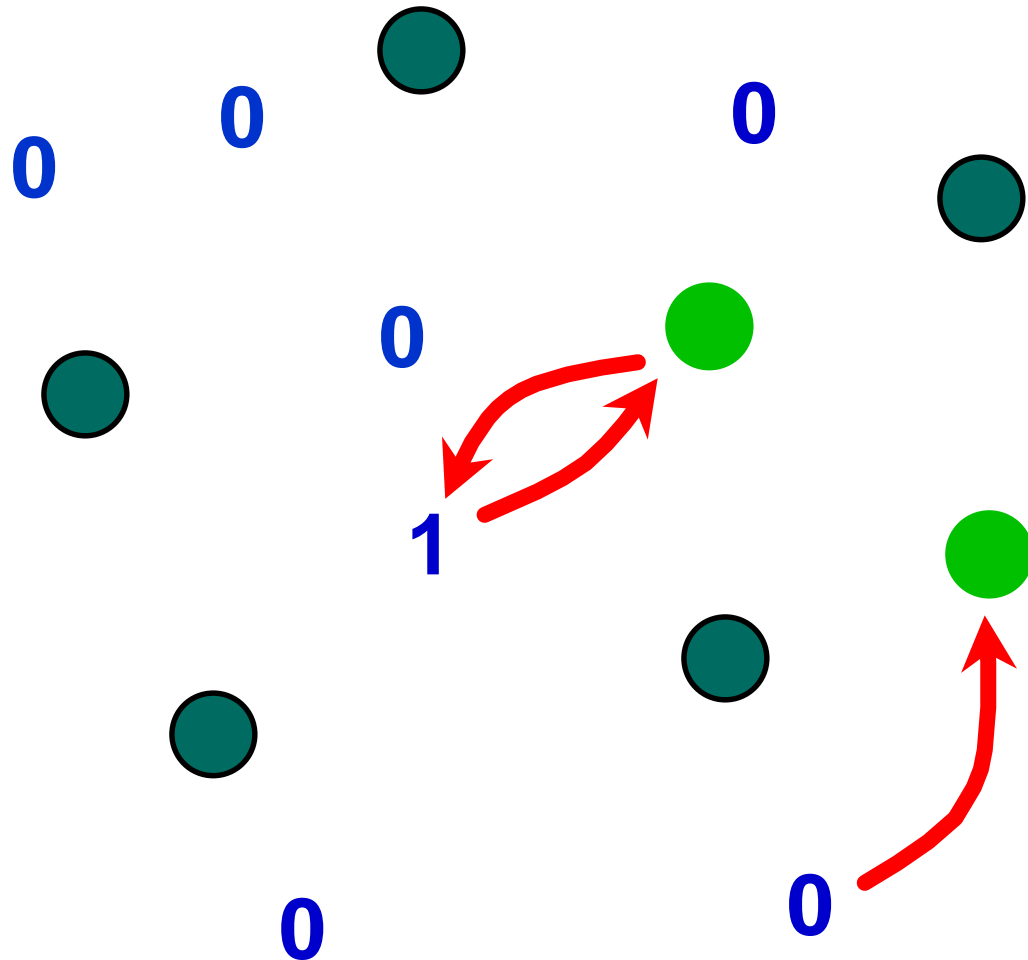
Selecting the coarse-grid points



Selecting the coarse-grid points

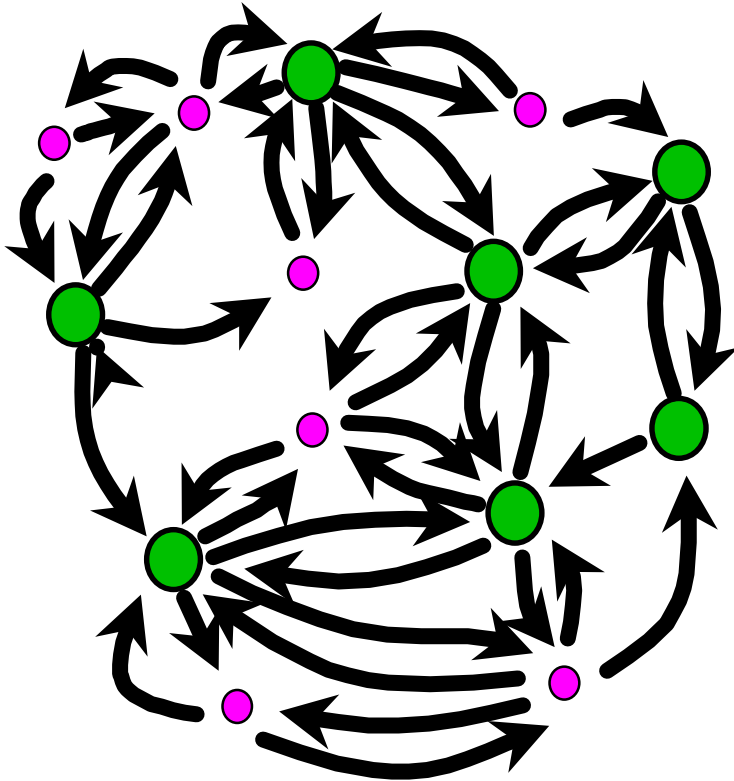


Selecting the coarse-grid points

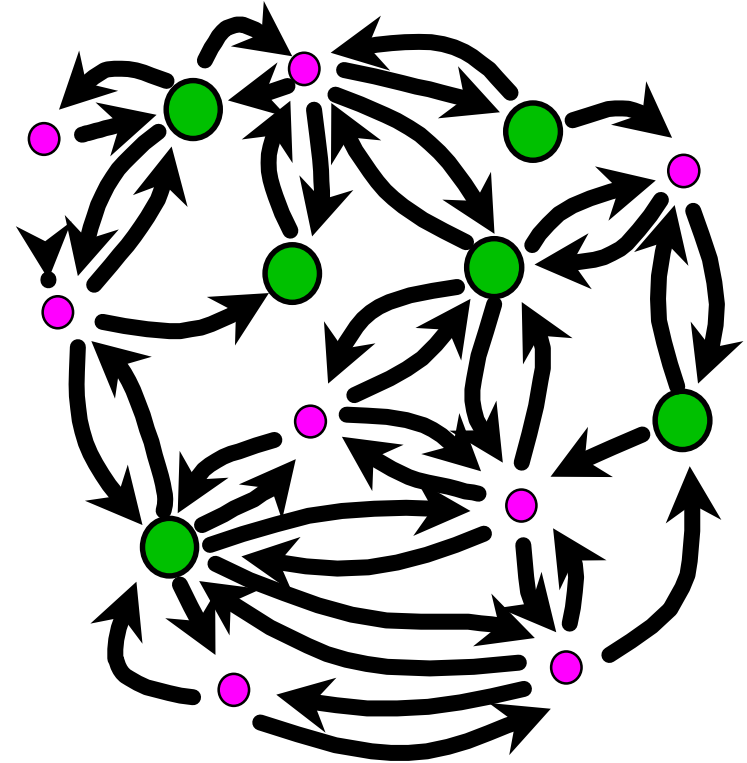


Select a final
independent set
and perform the
heuristics.

Selecting the coarse-grid points

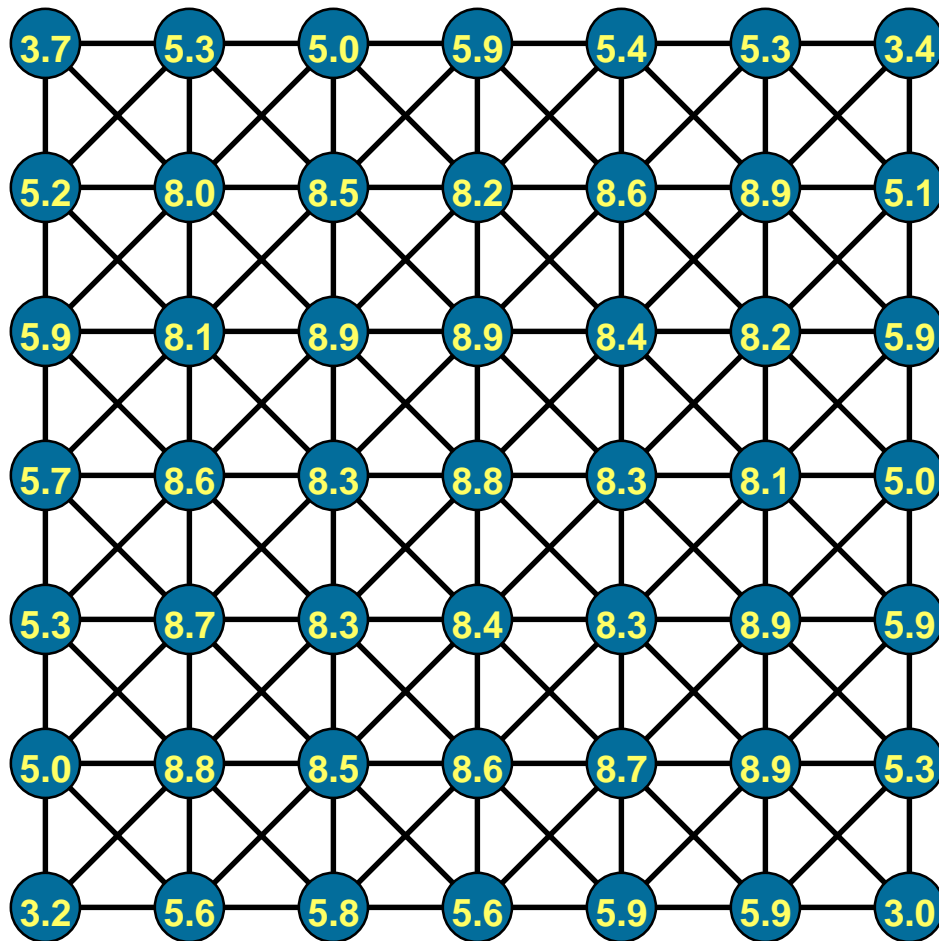


PAMG coarsening:
7 C-points selected



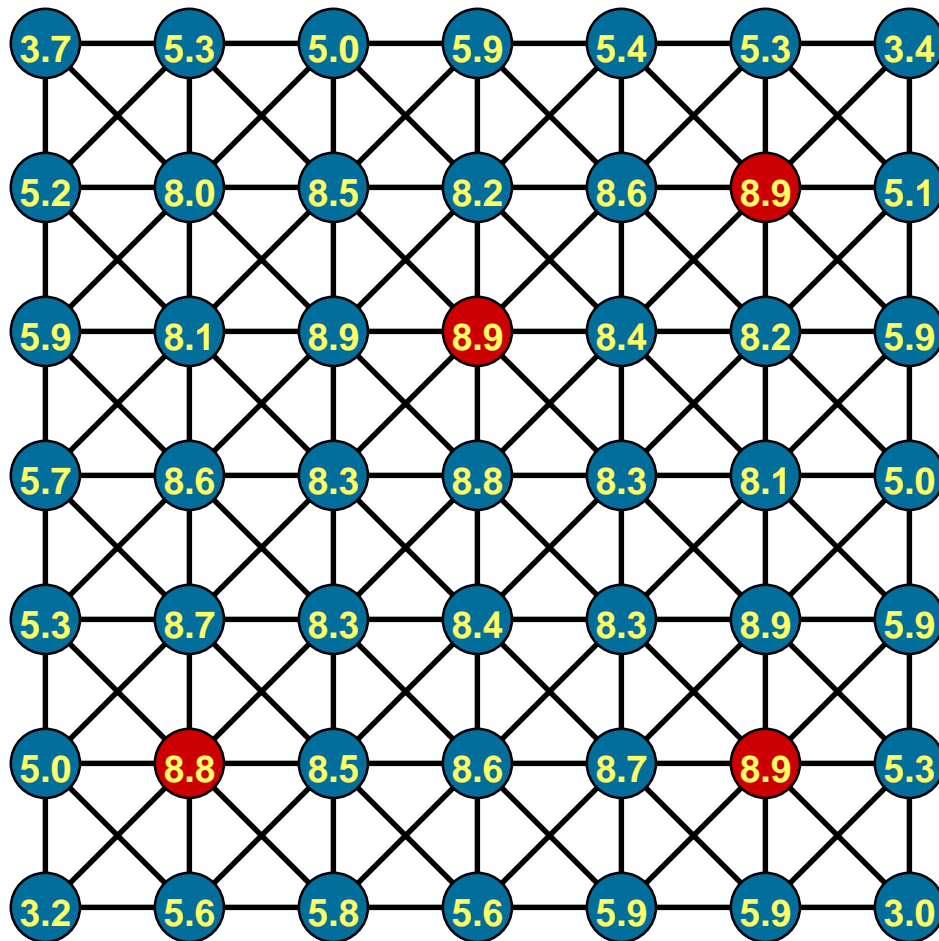
Standard AMG coarsening:
6 C-points selected

ParAMG start



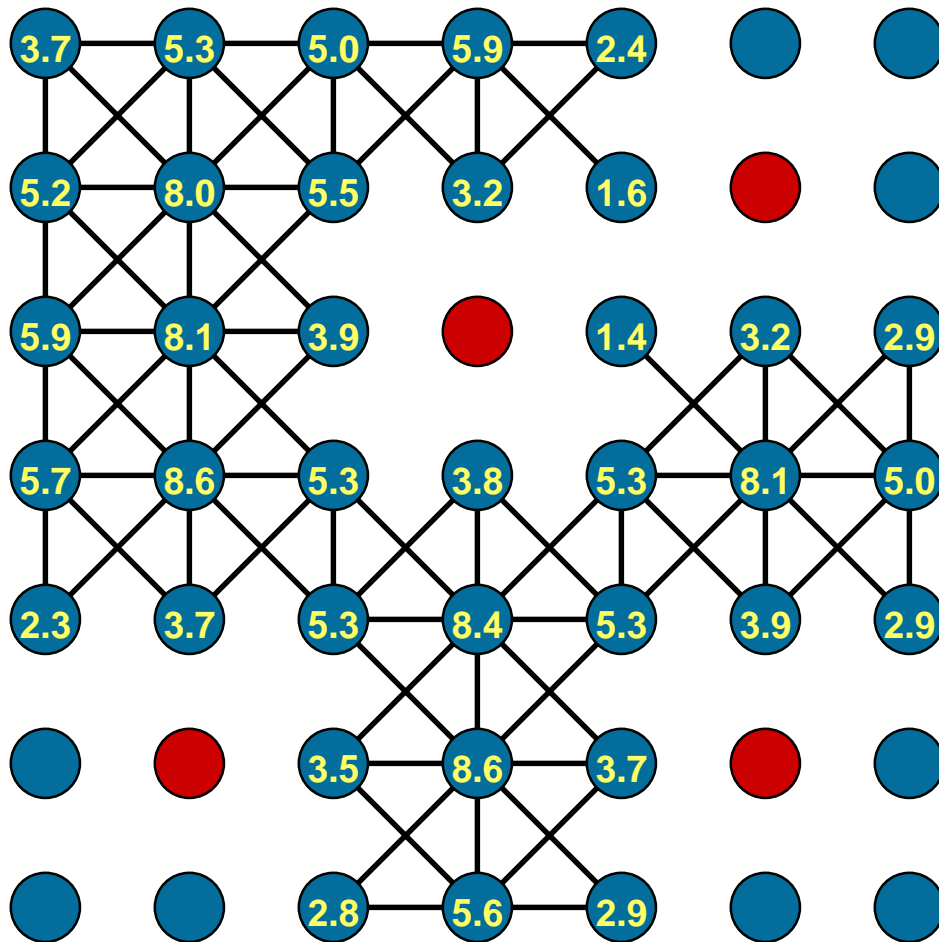
- ➡ select C-pts with maximal measure locally
- ➡ remove neighbor edges
- ➡ update neighbor measures

ParAMG select 1



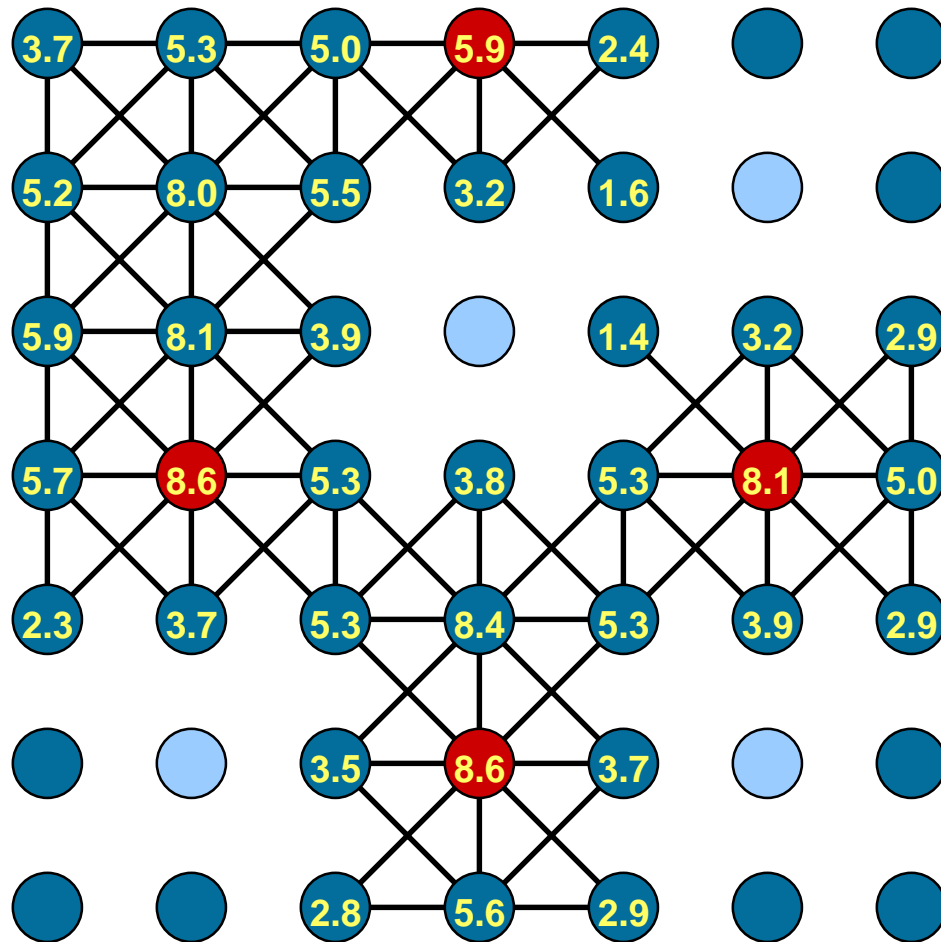
- ➡ **select C-pts with maximal measure locally**
- ➡ **remove neighbor edges**
- ➡ **update neighbor measures**

ParAMG remove and update 1



- ➡ select C-pts with maximal measure locally
- ➡ remove neighbor edges
- ➡ update neighbor measures

ParAMG select 2

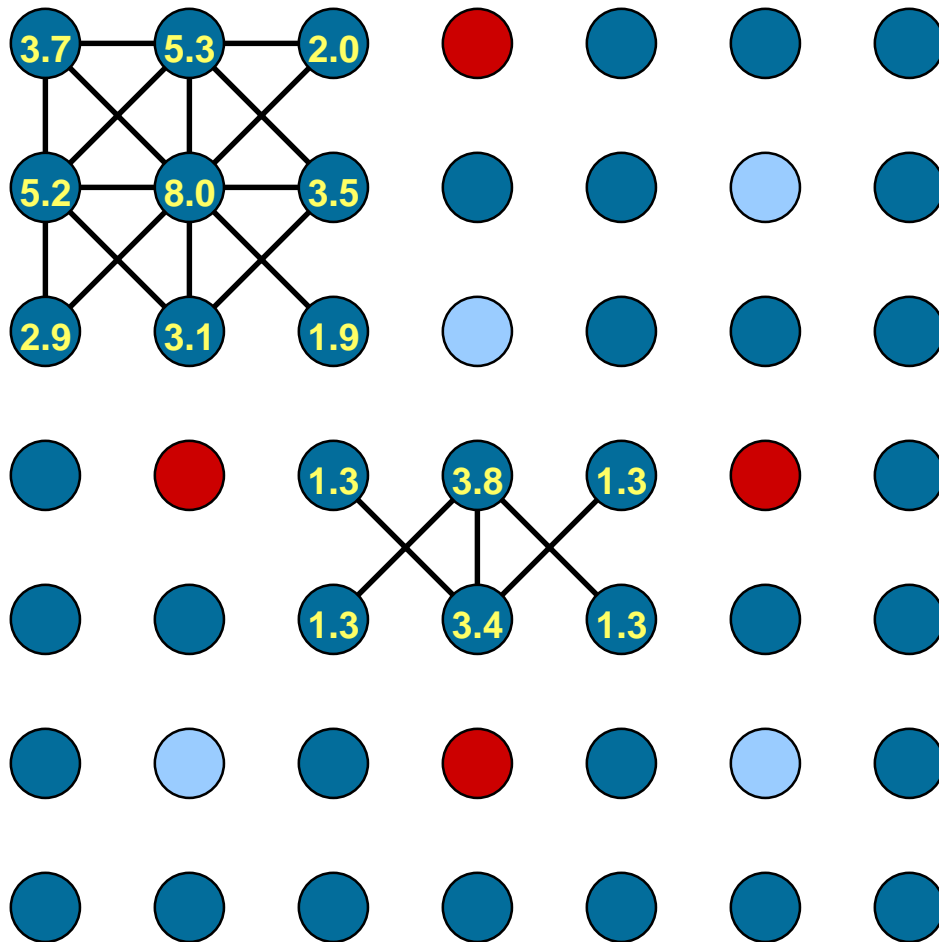


➡ **select C-pts with maximal measure locally**

➡ **remove neighbor edges**

➡ **update neighbor measures**

ParAMG remove and update 2

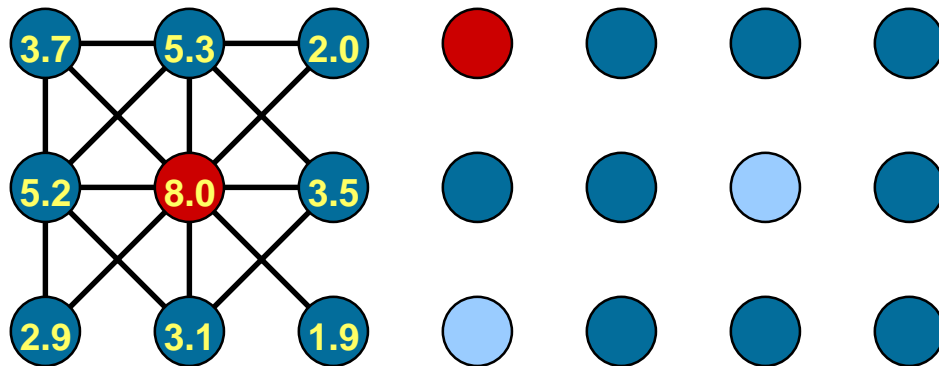


→ select C-pts with maximal measure locally

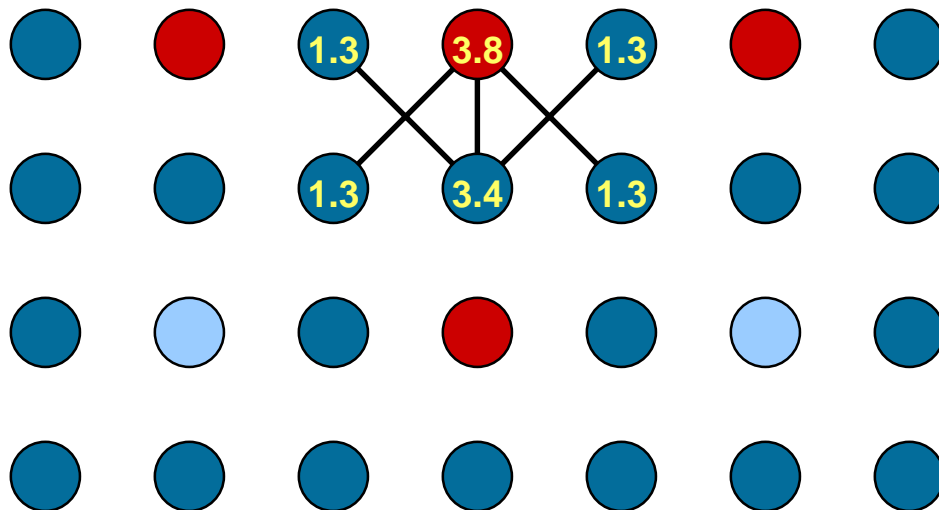
→ remove neighbor edges

→ update neighbor measures

ParAMG select 3



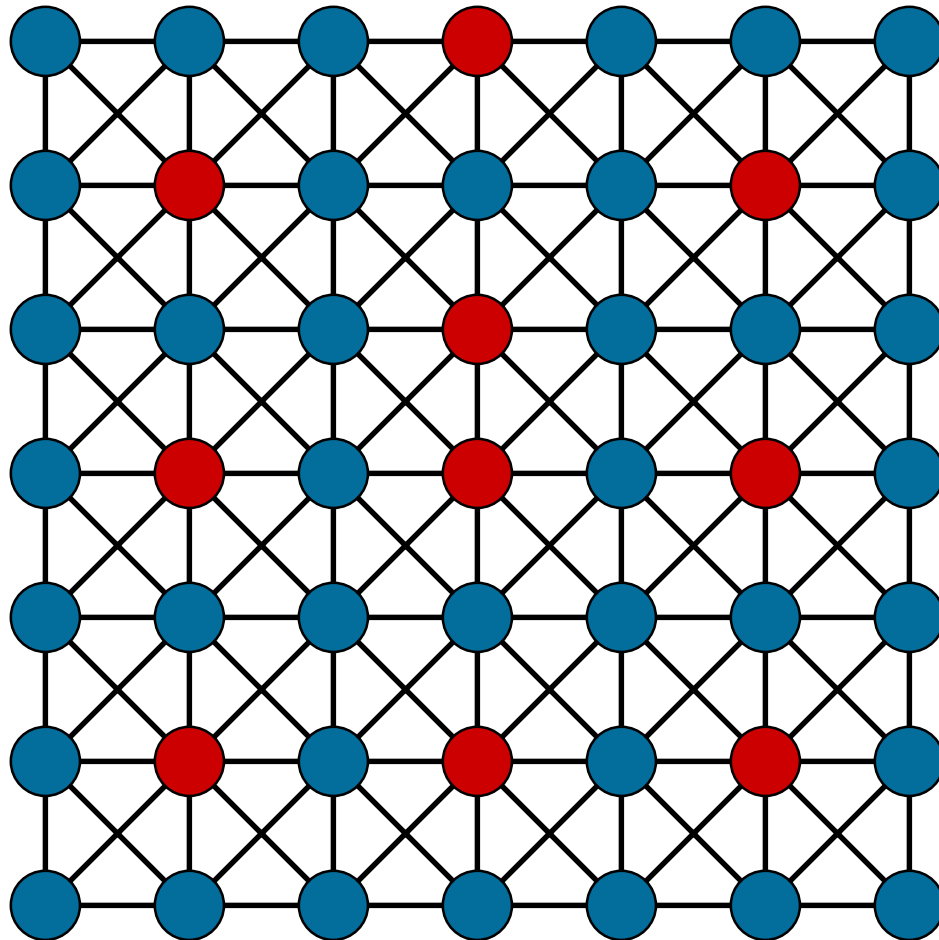
➡ **select C-pts with maximal measure locally**



➡ **remove neighbor edges**

➡ **update neighbor measures**

ParAMG final grid

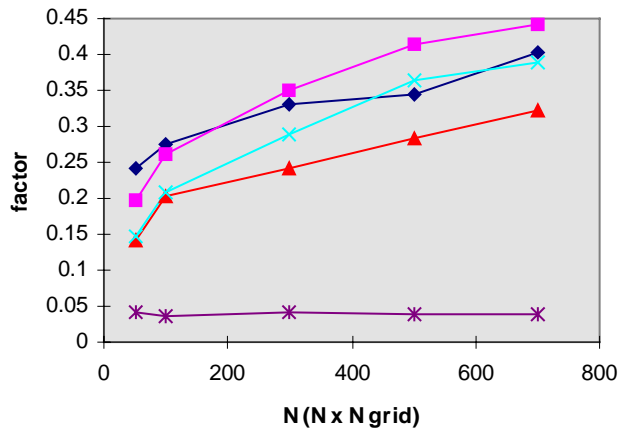


➔ **11 C-points
selected**

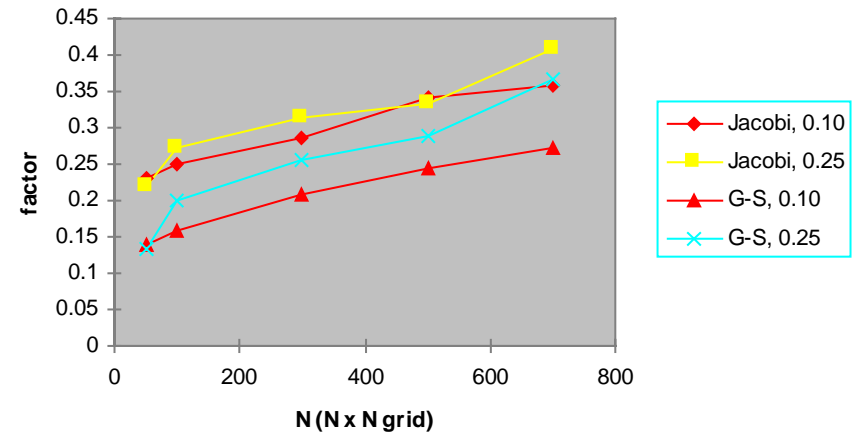
➔ **Standard AMG
selects 9 C-points**

ParAMG results: convergence factor

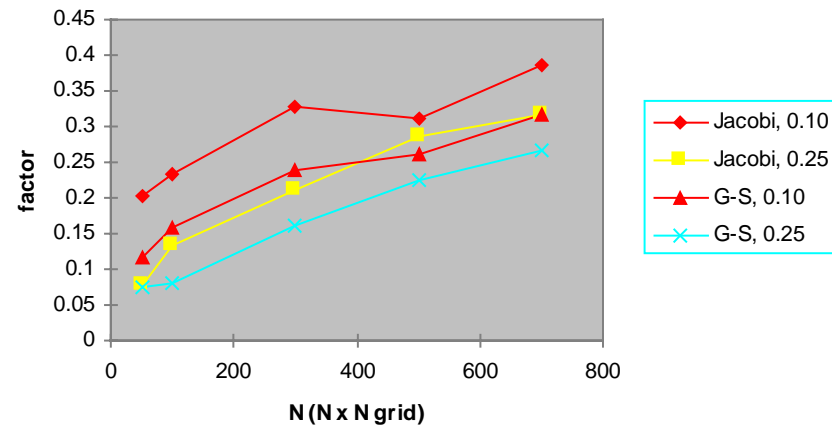
5-pt Laplacian: convergence



9-pt Laplacian: convergence

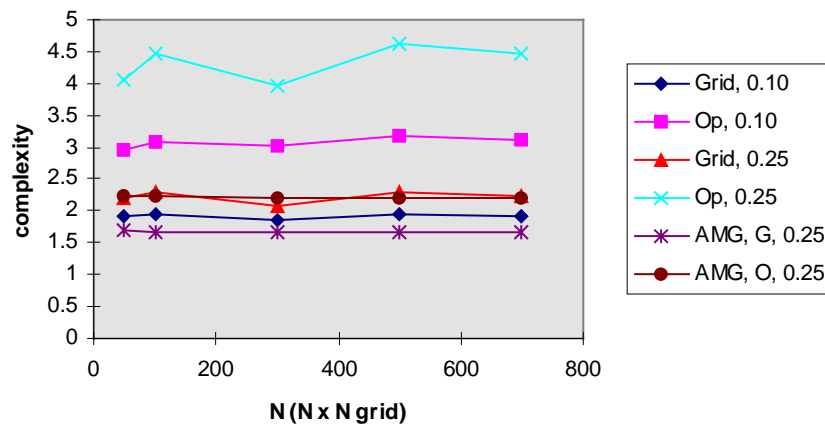


5-pt Anisotropic Laplacian: convergence

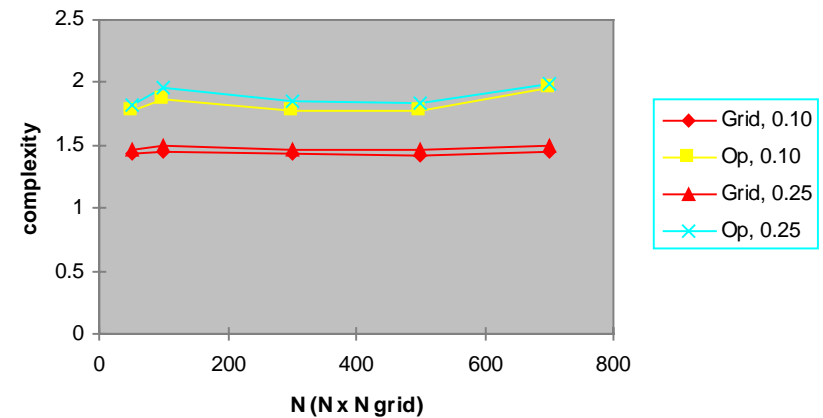


ParAMG results: complexity

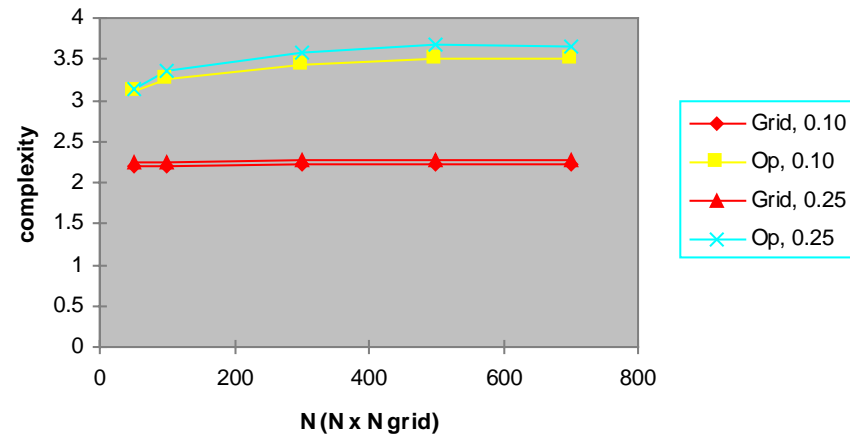
5-pt Laplacian: complexity



9-pt Laplacian: complexity



5-pt Anisotropic Laplacian: complexity



Back to AMG: what **else** can go wrong?

Thin body elasticity!

- **Elasticity, 3-d, thin bodies!**

$$u_{xx} + \frac{1-\nu}{2}(u_{yy} + u_{zz}) + \frac{1+\nu}{2}(v_{xy} + w_{xz}) = f_1$$

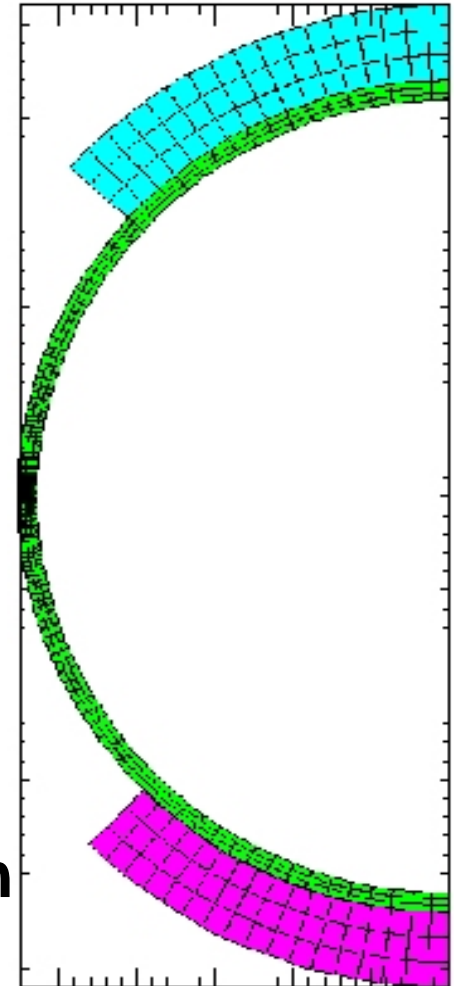
$$v_{yy} + \frac{1-\nu}{2}(v_{xx} + v_{zz}) + \frac{1+\nu}{2}(u_{xy} + w_{yz}) = f_2$$

$$w_{zz} + \frac{1-\nu}{2}(w_{xx} + w_{yy}) + \frac{1+\nu}{2}(u_{xz} + v_{yz}) = f_3$$

- **Slide surfaces, Lagrange multipliers, force balance constraints:**

$$\begin{pmatrix} S & T \\ U & V \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} F_1 \\ F_2 \end{pmatrix}$$

- **S is “generally” positive definite, V can be zero, $U^T \neq T$.**



We need a more robust characterization of smooth error

- Example: consider quadrilateral finite elements on a stretched 2D Cartesian grid ($\Delta x \rightarrow \infty$)

$$A = \begin{bmatrix} -1 & -4 & -1 \\ 2 & 8 & 2 \\ -1 & -4 & -1 \end{bmatrix}$$

- **Strong dependence is not apparent here**
- Iterative weight interpolation will sometimes compensate for mis-identified strong dependence
- Elasticity problems are still problematic

Global measures that relate interpolation accuracy and eigenmodes

- **Fundamental heuristic:** for a two grid algorithm, the interpolation operator must be able to reproduce a mode up to the same accuracy as the size of the associated eigenvalue.
- That is, one the following should be small

measure 1:
$$\frac{\langle (I - E) e, (I - E) e \rangle}{\langle Ae, e \rangle}$$

measure 2:
$$\frac{\langle A(I - E) e, (I - E) e \rangle}{\langle Ae, Ae \rangle}$$

Two-level convergence results

- If M is an upper bound for either measure 1 or 2, then the $V(1,0)$ convergence factor is bounded by

$$\alpha \leq \left(1 - \frac{1}{M \|A\|} \right)^{1/2}$$

- The $V(1,1)$ convergence factor is the square of above
- A multi-level result for measure 2 can be found in McCormick, SINUM 1985.

AMGe uses f.e. stiffness matrices to characterize smooth error locally

- For each point i , sum the local f.e. stiffness matrices

$$A_i = \sum_{k \in \eta_i} B_k$$

- We want the following local measures to be small

local measure 1:

$$\frac{\langle \varepsilon_i \varepsilon_i^T (I - E) e, (I - E) e \rangle}{\langle A_i e, e \rangle}$$

local measure 2:

$$\frac{\langle \varepsilon_i \varepsilon_i^T A_i \varepsilon_i \varepsilon_i^T (I - E) e, (I - E) e \rangle}{\langle A_i e, A_i e \rangle}$$

We use the local measures to define interpolation

- Given a coarse grid, we define interpolation to be the *arg min* of (for measure 1)

$$\min_{\substack{E}} \max_e \frac{\langle \epsilon_i \epsilon_i^T (I - E) e, (I - E) e \rangle}{\langle A_i e, e \rangle}$$

- Similarly for measure 2

Using local measures to define interp. is equivalent to fitting local eigenmodes

- Assume the eigen-decomposition:

$$A_i V = \Lambda V; \quad \Lambda = \begin{bmatrix} 0 & 0 \\ 0 & \Lambda_+ \end{bmatrix}; \quad V = \begin{bmatrix} V_0 & V_+ \end{bmatrix};$$

- Finding the *arg min* of measures 1 and 2 is equivalent to solving the following constrained least-squares problems with $p=1/2$ and $p=1$

$$\min_w \left\| \Lambda_+^{-p} V_+^T \begin{pmatrix} 1 \\ 0 \\ -w \end{pmatrix} \right\|^2, \quad \text{s.t.} \quad V_0^T \begin{pmatrix} 1 \\ 0 \\ -w \end{pmatrix} = 0$$

Computing interpolation in practice

- For measure 1, partition local matrix by F and C -pts:

$$A_i = \begin{bmatrix} A_{ff} & A_{fc} \\ A_{cf} & A_{cc} \end{bmatrix}$$

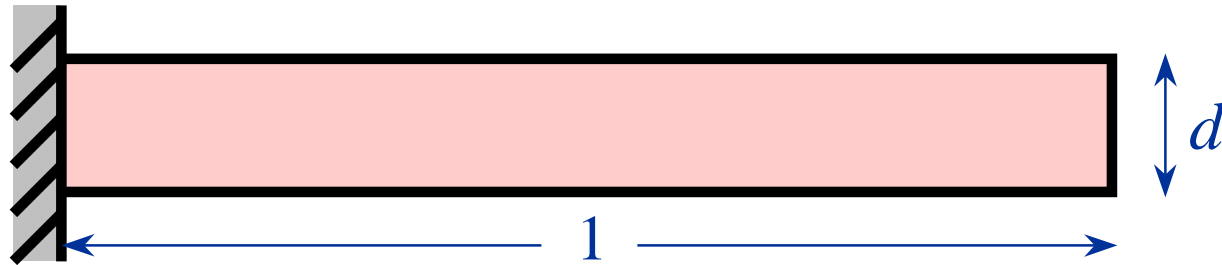
- Then interpolation to point i is defined by

$$w^T = \varepsilon_i^T W, \quad \text{s.t. } A_{ff} W = -A_{fc}$$

- Measure 1 seems to produce better interpolation in practice than does measure 2

Preliminary results for new AMGe prolongation are promising

- 2D plane-stress cantilever beam, fixed on one end.



- **Convergence factors:** finest grid elements are $h \times h$; grids are coarsened geometrically.

d	h	AMG prolongation	AMGe prolongation
1	1/32	0.60	0.20
1/4	1/8	0.95	0.25
1/8	1/16	0.90	0.26
1/16	1/64	0.92	0.26

Bounds on local measure 1 (and 2) yield global convergence bounds

$$\begin{aligned}\langle (I-E)e, (I-E)e \rangle &= \sum_{i \in F} \langle \varepsilon_i \varepsilon_i^T (I-E)e, (I-E)e \rangle \\ &\leq \sum_{i \in F} M_i \langle A_i e, e \rangle \\ &= \sum_k \langle B_k e, e \rangle \sum_{i \in N_k \cap F} M_i \\ &\leq M \langle A e, e \rangle\end{aligned}$$

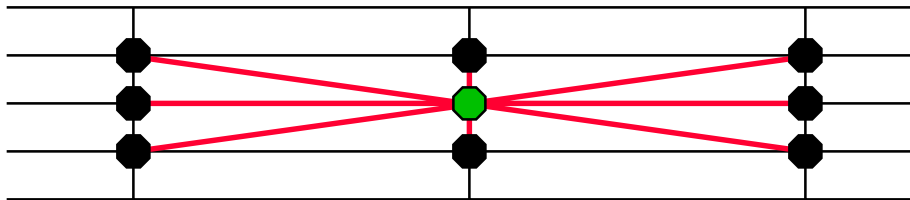
- We would like to use this bound to assess the quality of interpolation and coarse-grid choice

$$\alpha \leq \left(1 - \frac{1}{M \|A\|} \right)^{1/2}$$

Using the local measure to define “strong dependence”

- Consider interpolating to point i independently from each of its neighbors:

For each point k connected to point i , compute value of M_i assuming k is the only C-point.



$$A = \begin{bmatrix} -1 & -4 & -1 \\ 2 & 8 & 2 \\ -1 & -4 & -1 \end{bmatrix}$$

M_i values for quadrilateral
example (1000:1)

$$\begin{matrix} 3 \times 10^6 & 9.7 & 3 \times 10^6 \\ 3 \times 10^6 & & 3 \times 10^6 \\ 3 \times 10^6 & 9.7 & 3 \times 10^6 \end{matrix}$$

Target Problems:

- **Adaptive, Lagrangian-Eulerian 3d code :**
 - Radiation-Hydro, manufacturing, gross deformation
 - slide surfaces & thin bodies
- **We have found that AMG-preconditioned GMRES can be used effectively on several typical problems**
- **We hope to test AMGe ideas out on several test problems soon**
 - now able to get stiffness matrices
 - will use a modified AMG coarsening heuristic to coarsen appropriately near boundaries
 - look first at a problem without slide surfaces

Conclusions

- *AMG has been shown to be a robust, efficient solver on a wide variety of problems of real-world interest.*
- **AMG can be parallelized using a modified Luby-Jones-Plassman parallel MIS algorithm to develop the parallel coarse-grid selection.**
- **Tests show that the parallel algorithm will produce grids that give acceptable (not optimal) convergence factors. Future work includes modifying the coarsening algorithm to improve performance and ensure scalability.**
- **Finite element stiffness matrices and local measures can be used to better characterize smooth error, select coarse grids, and construct interpolation.**

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